



# CLIFDEW-GRID: Early Warning Grid-Based Risk Modelling of Climate Induced Forced Displacement

Climate and displacement workshop:  
04 - 05 November 2025

Data Science Team  
Statistics, Data Science, and Survey Section  
Global Data Service



# Background

# Background: project overview

## Aims and Goals



### Aim

#### Risk index for forced displacement

- Subnational (grid cell) level
- Monthly frequency

### Case area

East, Central, and West Africa

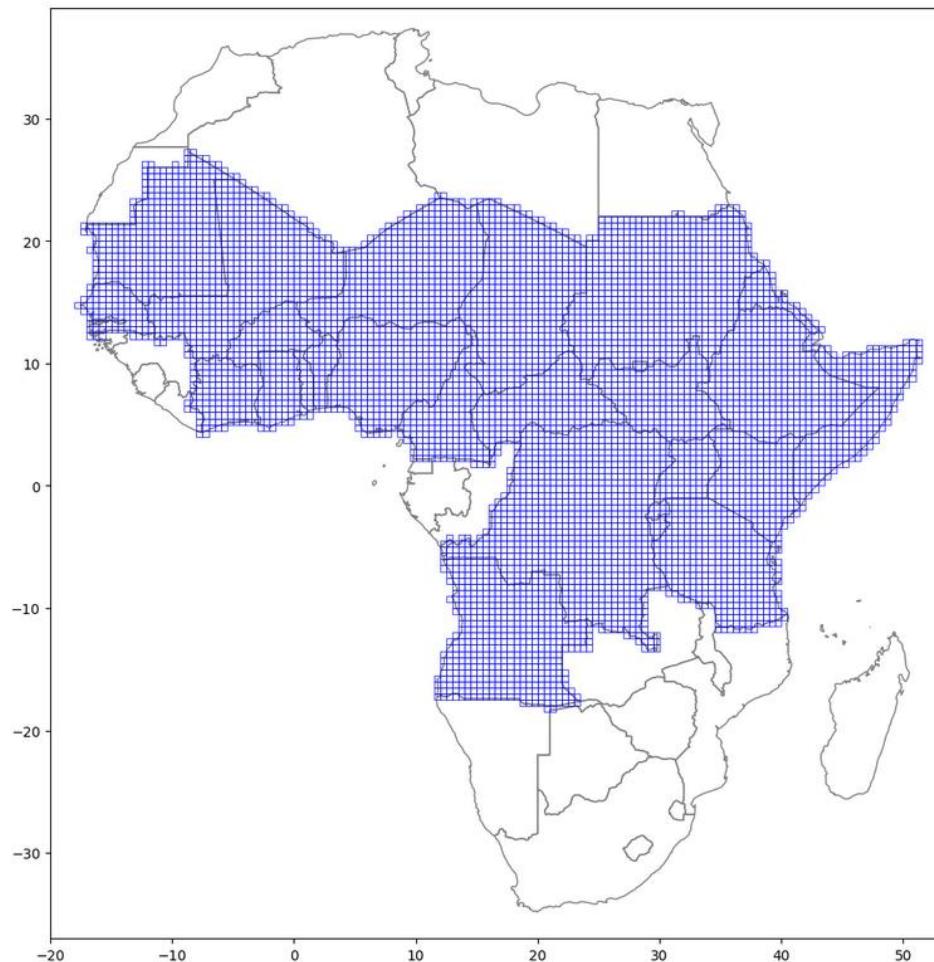
### Goal

#### Early Warning Model and Insights

- Provide UNHCR and other humanitarian agencies and stakeholders with a tool for anticipatory action relating to forced displacement, taking climate change into account.
- Create deeper insights into the nexus between climate change and forced displacement.

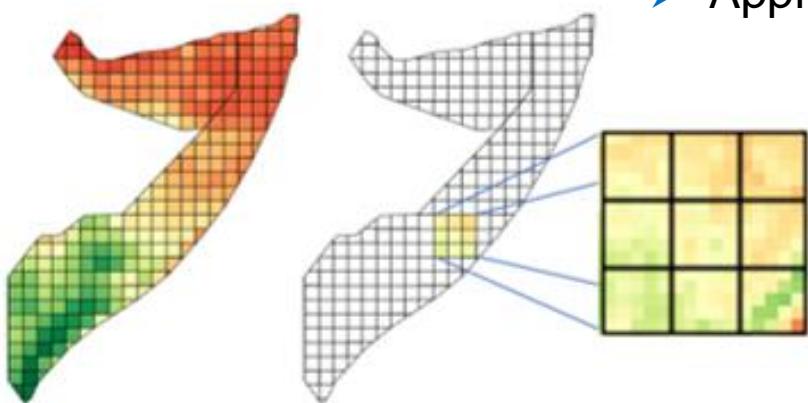
# Geographical focus

- East, Central and West Africa
- $0.5^\circ$  grid-cells (approx. 55 km x 55km)
- Approx. 6220 grid-cells

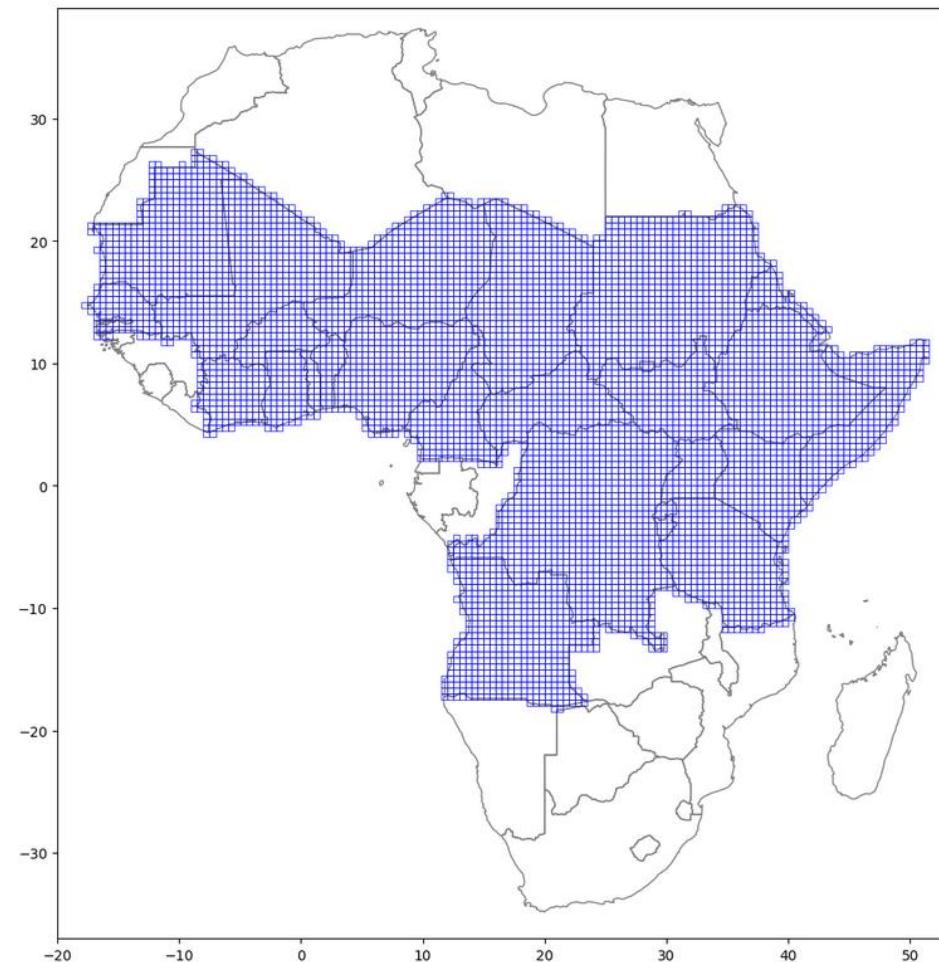


# Geographical focus

- East, Central and West Africa
- $0.5^\circ$  grid-cells (approx. 55 km x 55km)
- Approx. 6220 grid-cells



- Feature variable data collected at the  $0.1^\circ$  grid-cell level to account for variance
- Approx. 150,000 grid cells

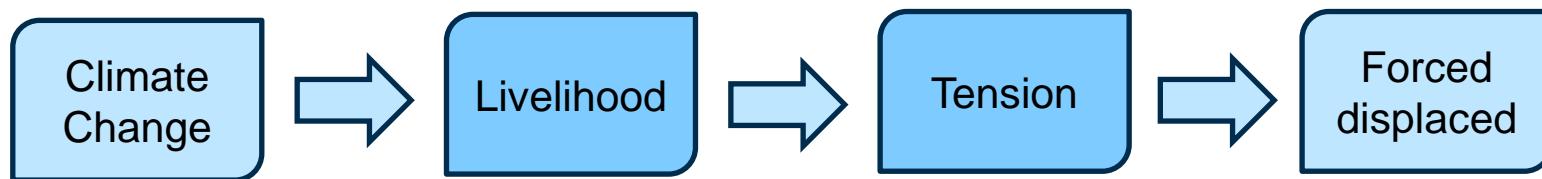


# Project structure

# Hypothesis

## Modelling structure

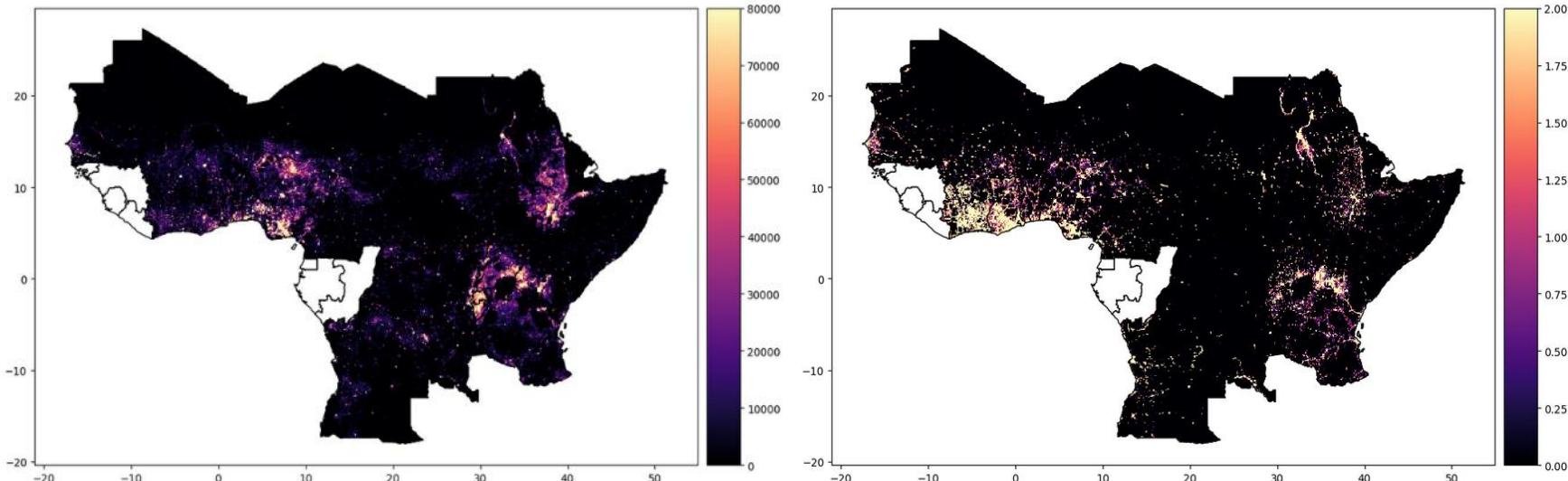
- »» Slow on-set climate change generates loss of livelihood.
- »» Loss of livelihood can lead to competition over diminishing resources, resulting in political tension.
- »» Conflict contributing to forced displacement.



# Intermediate models

# Monthly population

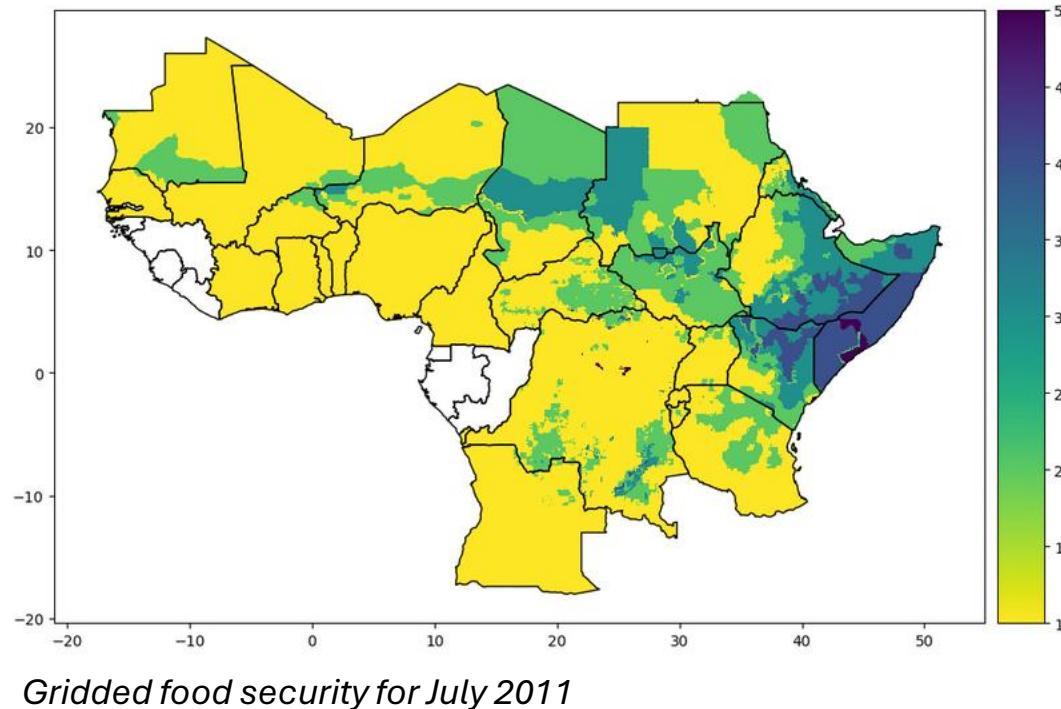
- LandScan population data
  - Annual frequency
  - 1000-meter resolution
- Based on sub-national information, observations are placed inside  $0.1^\circ$  grid-cells and aggregated to monthly time intervals using smoothing method
- Use monthly nightlight data (Defense Meteorological Satellite (DMSP) and Visible Infrared Imaging Radiometer Suite (VIIRS)) to correct monthly trends



Gridding LandScan population figures in year 2020 (left), gridded VIIRS nightlight radiance in May 2020 (right)

# Monthly food security

- Famine Early Warning System Network (FEWS NET)
  - Food security evaluation within subnational geographical areas
  - Released every 3 months with current situation, 1 – 3 month prediction and 4 – 6 month prediction
- Backcasting model to predict monthly food security for time period before historical data is available
  - LightGBM with series of feature variables
- Apply forecasting model when waiting for new data to be released



# Gridding ProGres Displacement



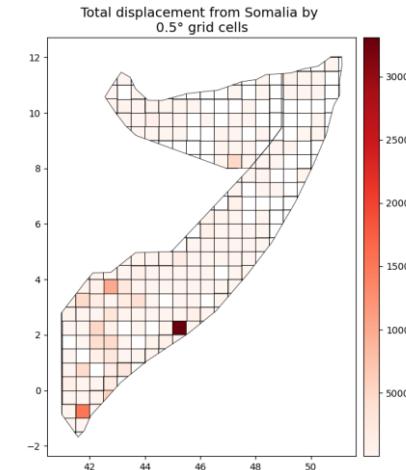
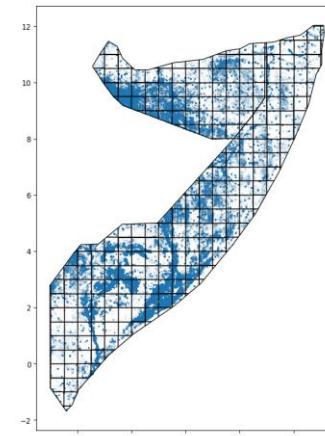
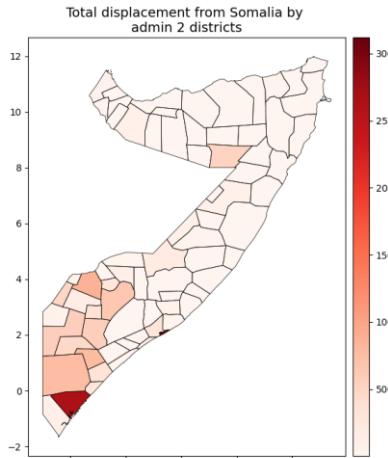
Success of the CLIFDEW project will hinge on UNHCR's ProGres data and especially placing observations within grid-cells



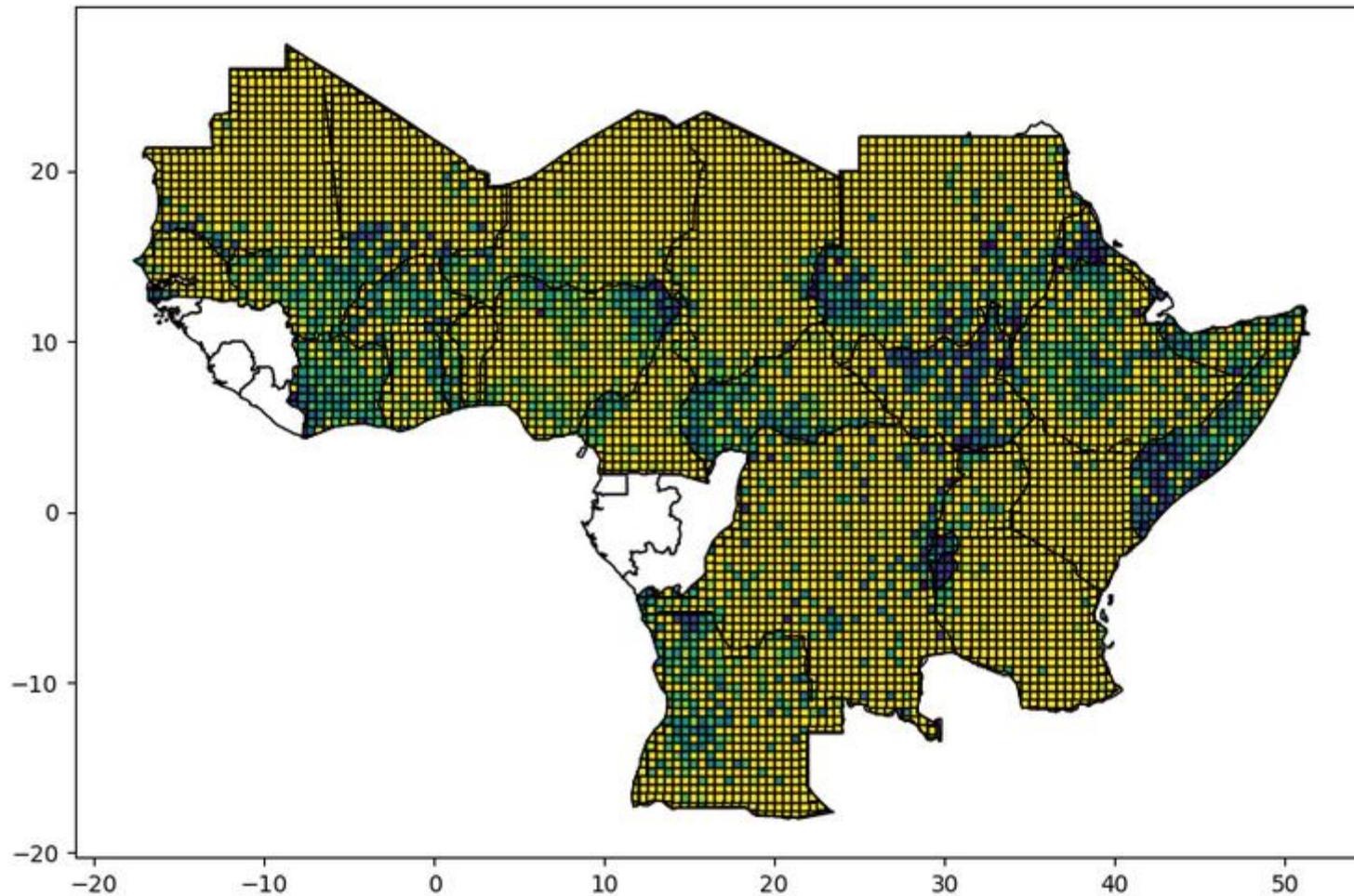
Need a method for placing observations into grid cells based on known subnational locations in data



Predict grid-cell locations for observations using building prior key and other variables



# Gridding ProGres Displacement



# Feature variables

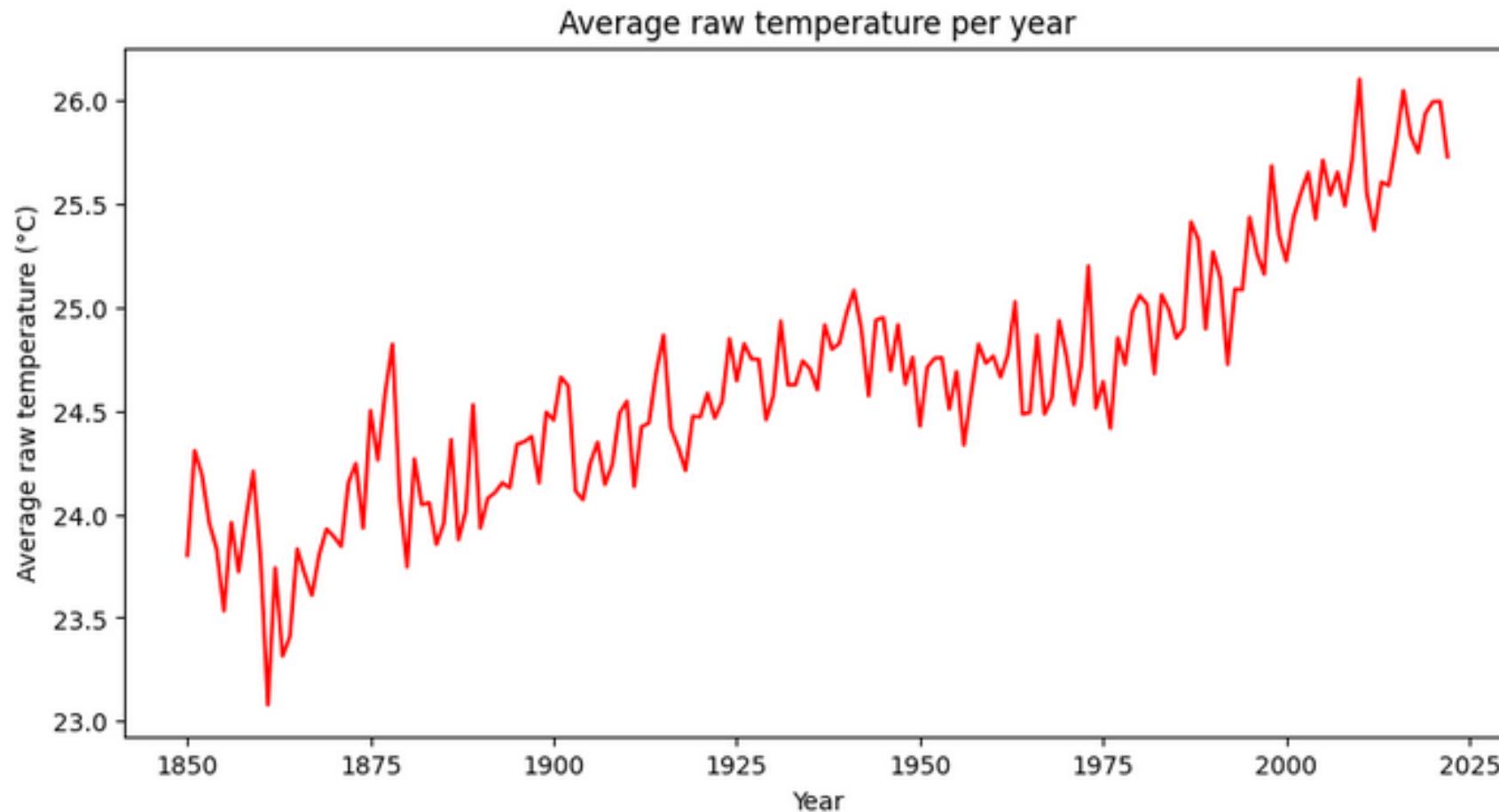
Variable	Unit	Source
<b>Climate variables</b>		
Temperature	Daily, 0.05° Daily, 0.10° Monthly, 0.25°	CHIRTS Copernicus ERA-5 Berkeley Earth
Precipitation	Daily, 0.05° Daily, 0.10°	CHIRPS Copernicus ERA-5
Normalized Difference Vegetation Index	Monthly, 0.05°	NASA
Standardized Precipitation Evapotranspiration Index	Monthly, 1.00°	SPEI Global Drought Monitor

## Resource and geographic variables

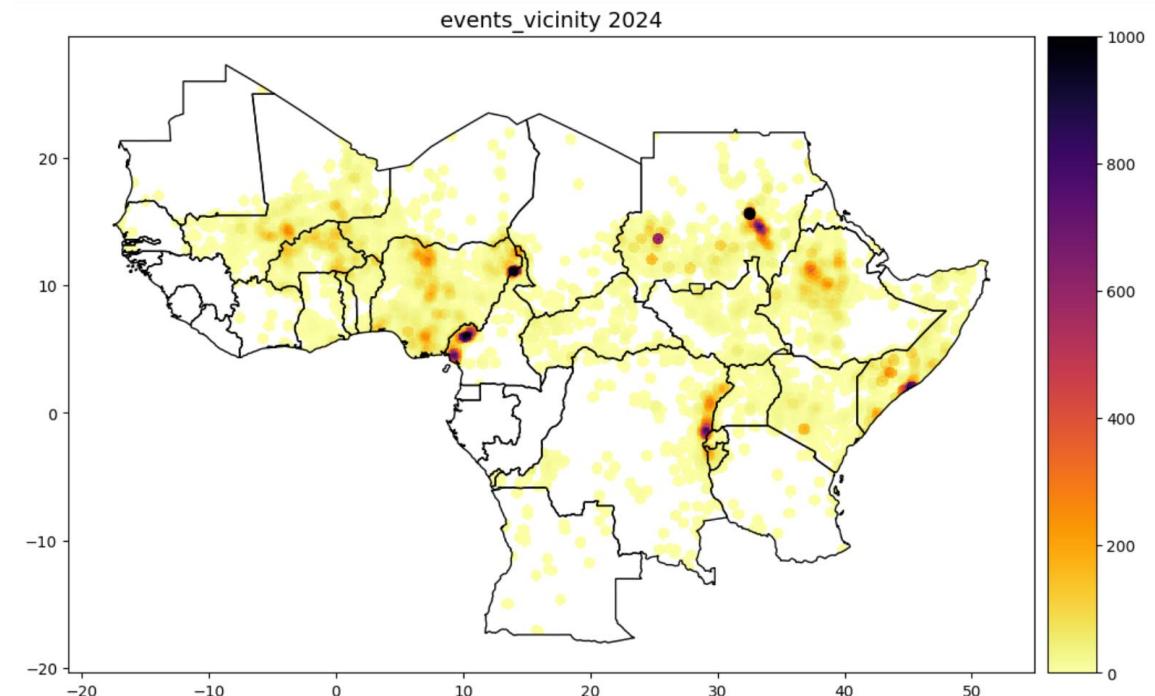
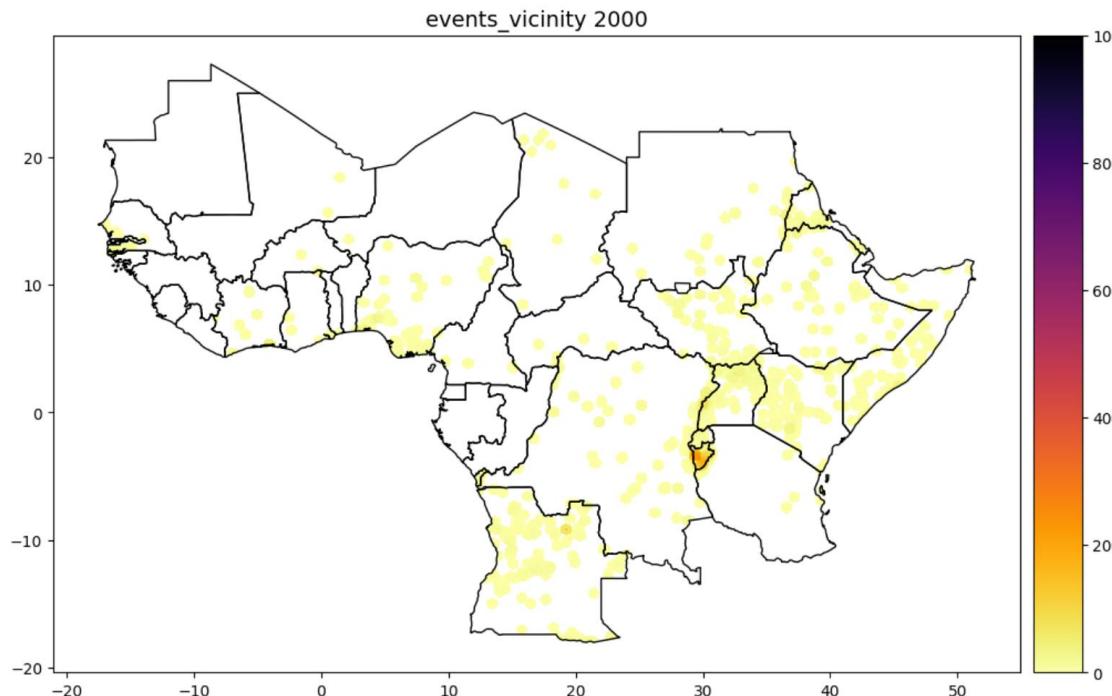
Resource and geographic variables	Source	Notes
Landcover	Stagnant, 0.05°	Copernicus
Agro-ecological zone	Stagnant, defined regions	International Food Policy Research Institute
Elevation	Stagnant, 0.10°	HarvestChoice CELL5M
River	Stagnant, defined regions	Natural Earth
Road	Stagnant, defined regions	Humanitarian OpenStreetMap
Market access	Stagnant, 0.10°	International Food Policy Research Institute
Subsistence Index	Stagnant, 0.10°	International Food Policy Research Institute, Harvard Dataverse

<b>Demographic variables</b>		
Population density	Monthly, 0.10°	LandScan, DMSP, VIIRS
Ethnicity	Stagnant, 0.10°	ETH Zurich
<b>Food security</b>		
Predicted food security	Monthly, 0.10°	FEWS NET
<b>Socio-economic and wellbeing variables</b>		
Child health	Monthly, 0.50°	PRIOR
Gini	Monthly, 0.10°	WorldPop, VIIRS
<b>Political variables</b>		
Fragility	Annual, national	Fragile States Index
<b>Conflict variables</b>		
Conflict	Daily, geo-point locations	ACLED

# Climate conditions: temperature over time



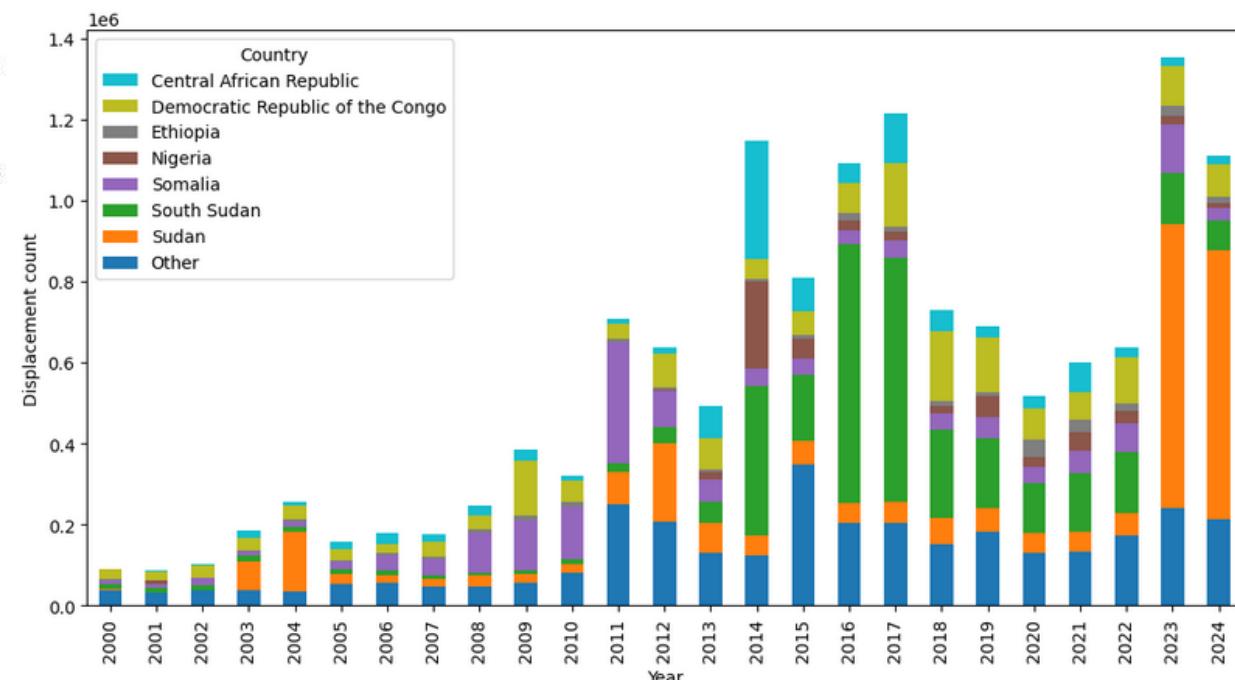
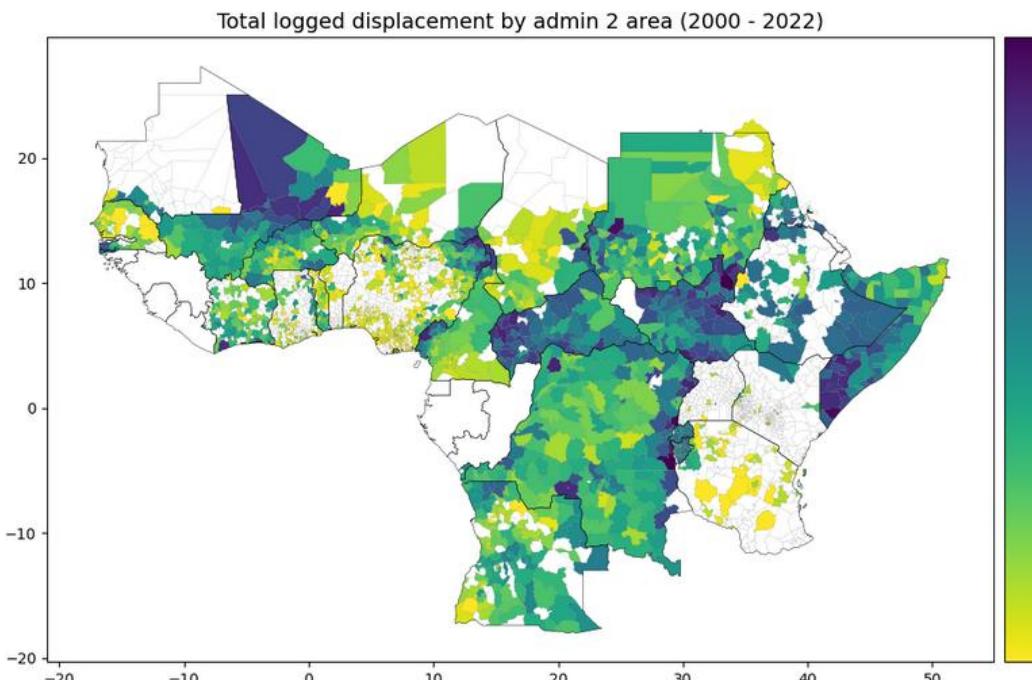
# Tension: conflict



# Target variable

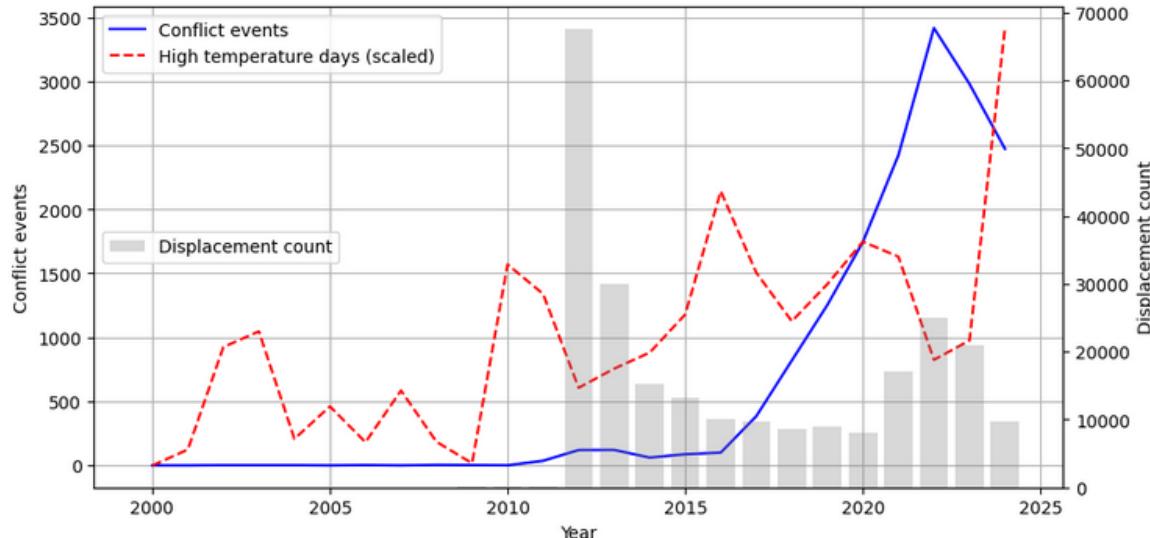
# Target: displacement

- UNHCR ProGres data
  - Registry of asylum seekers and refugees
  - Approximately 18 million observations, 13.9 million relevant to this project
- Based on sub-national information, observations are placed inside  $0.5^\circ$  grid-cells and aggregated to monthly time intervals

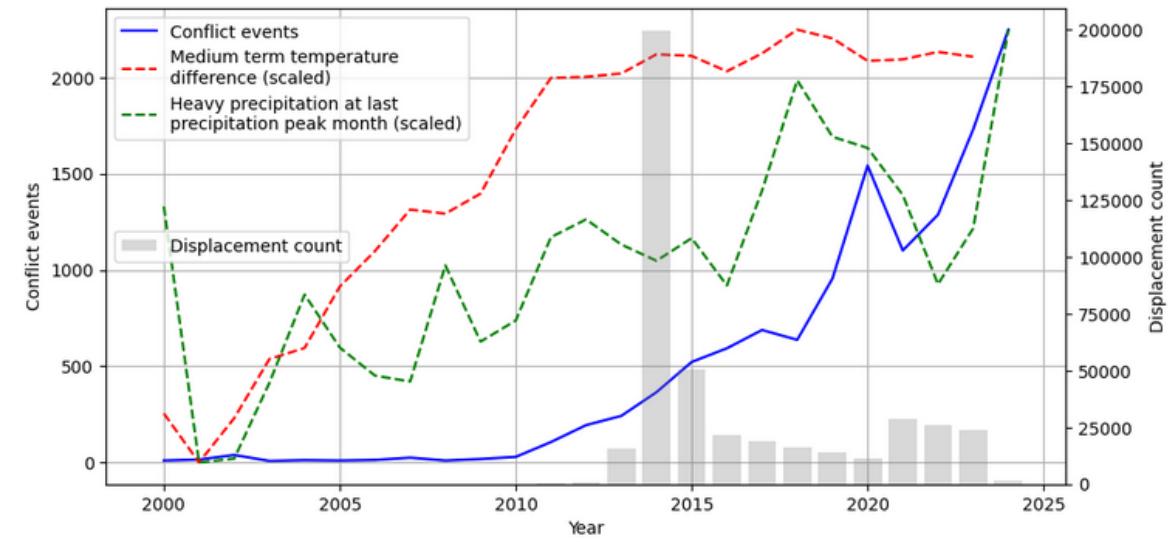


# Target: displacement

Liptako-Gourma (Mali, Burkina Faso, Niger)



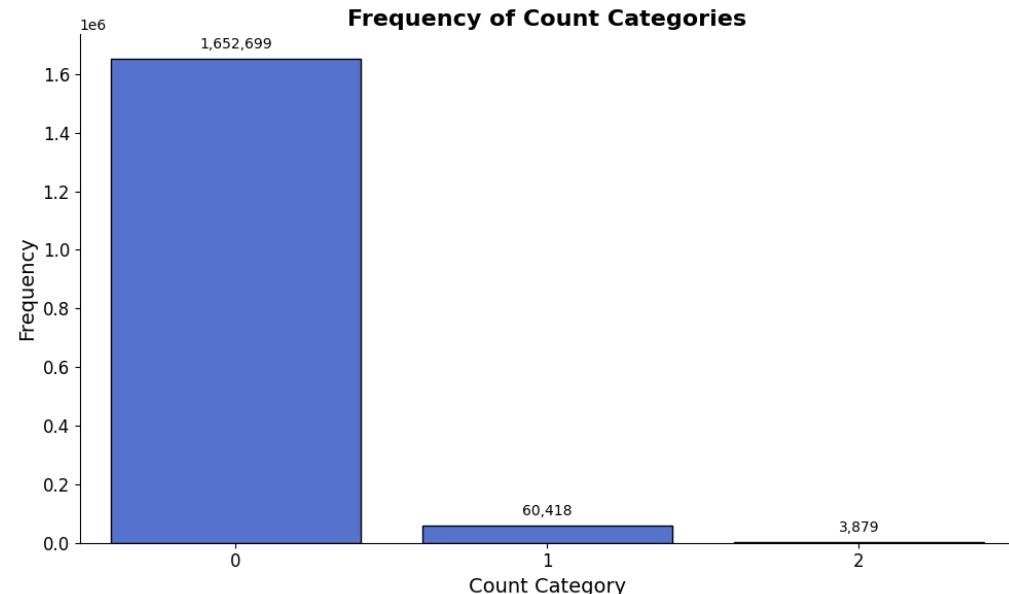
## Lake Chad



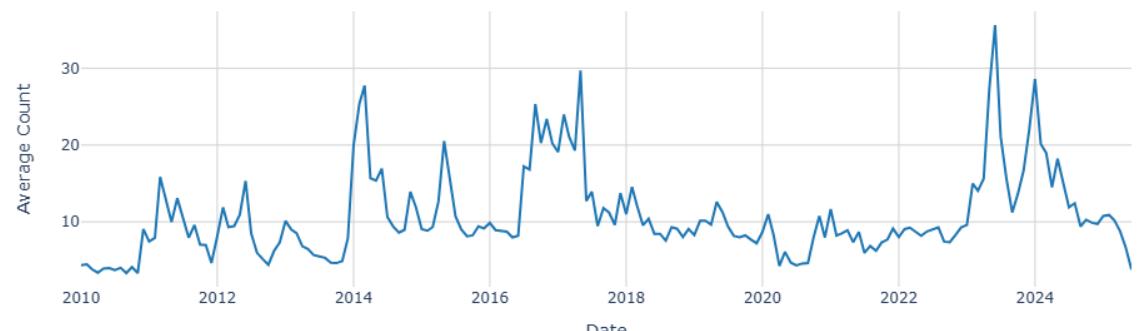
# Predicting Forced Displacement Risk

# Response variable

- Cross-border movements, no IDPs (UNHCR data)
  - 1,757 of 6225 grid cells (28.2%) have experienced some displacement
- We have created three levels of forced displacement:
  - Class 0, (**low**, 0-9 forcibly displaced people)
  - Class 1, (**medium**, 10-500)
  - Class 2, (**high**, >500)
- We are predicting the probability of one of these levels occurring for each grid
  - we are predicting the risk of people fleeing **from** a grid
- We predict this for 1, 3, and 6 months into the future



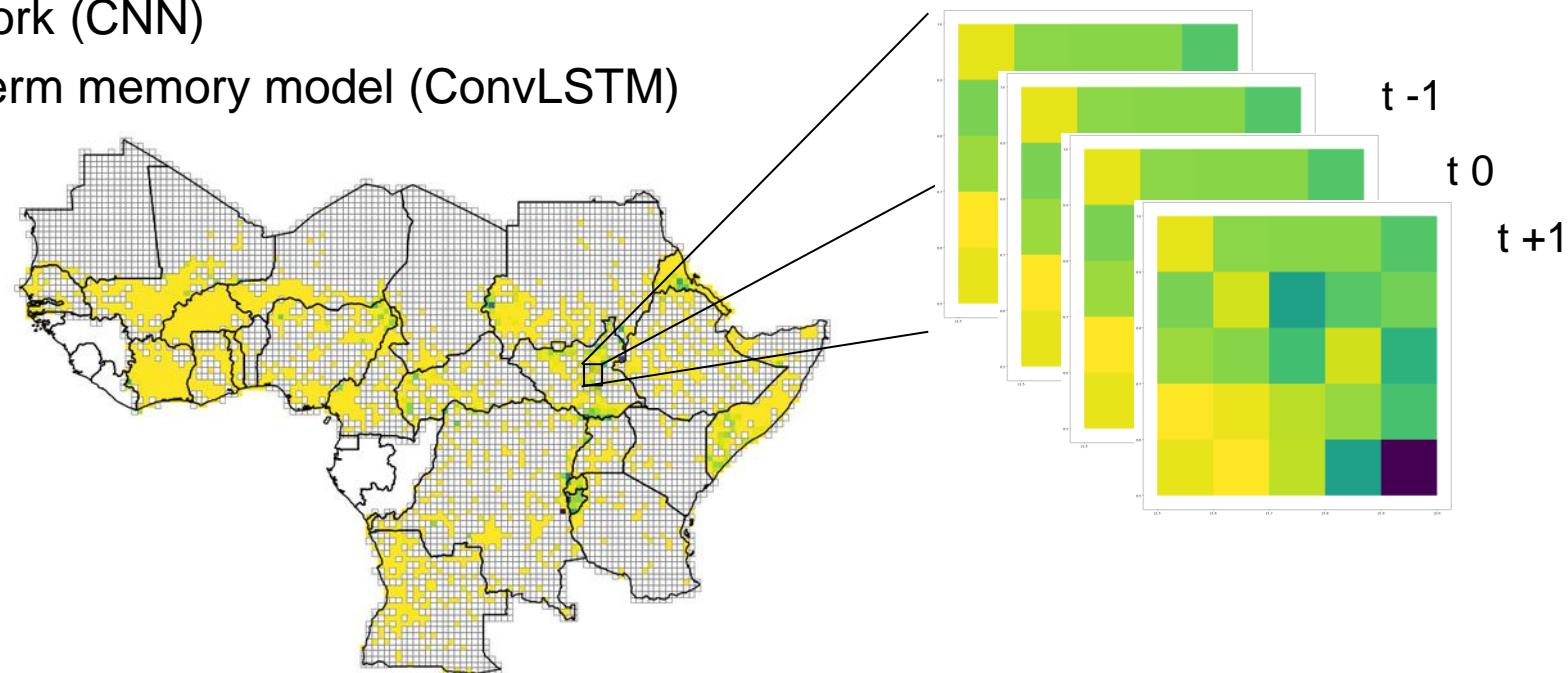
Average Count Across Grids Over Time



Average displacement per grid cell from 2000 to 2025.

# Models used

- We use an Ensemble model which combines different types of models
  - Baseline model
  - LightGBM model
  - Neural networks models
    - Convolutional neural network (CNN)
    - Convolutional long short-term memory model (ConvLSTM)
    - Hierarchical ConvLSTM
  - Ensemble models combines the strengths of all models to produce the most accurate predictions



# Model findings

- LightGBM model predicting class 2 (>500) 6 months ahead
- Top 10 climate variables ranked by the highest average impact on the model predictions

Number of months since precipitation peak

Long-term temp. anomaly from 1950-79 temp. baseline

Max. temp. anomaly for last 6 months from 1980-90 temp. baseline

Mean temp. anomaly for last 3 months from 1980-90 temp. baseline

Average max. precip. for last 6 months

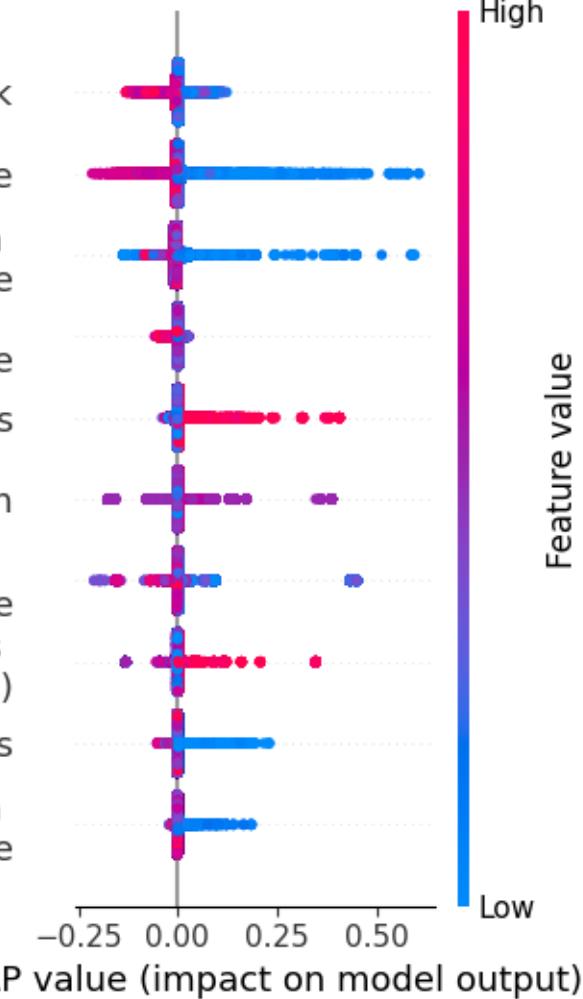
Precip. anomaly 4 months prior last precip. peak month

Max. precip. anomaly for last 6 months from 1980-90 precip. baseline

Num. of killing days anomaly for last 6 months from baseline (1980-1990)

Sum of drought level over last 12 months

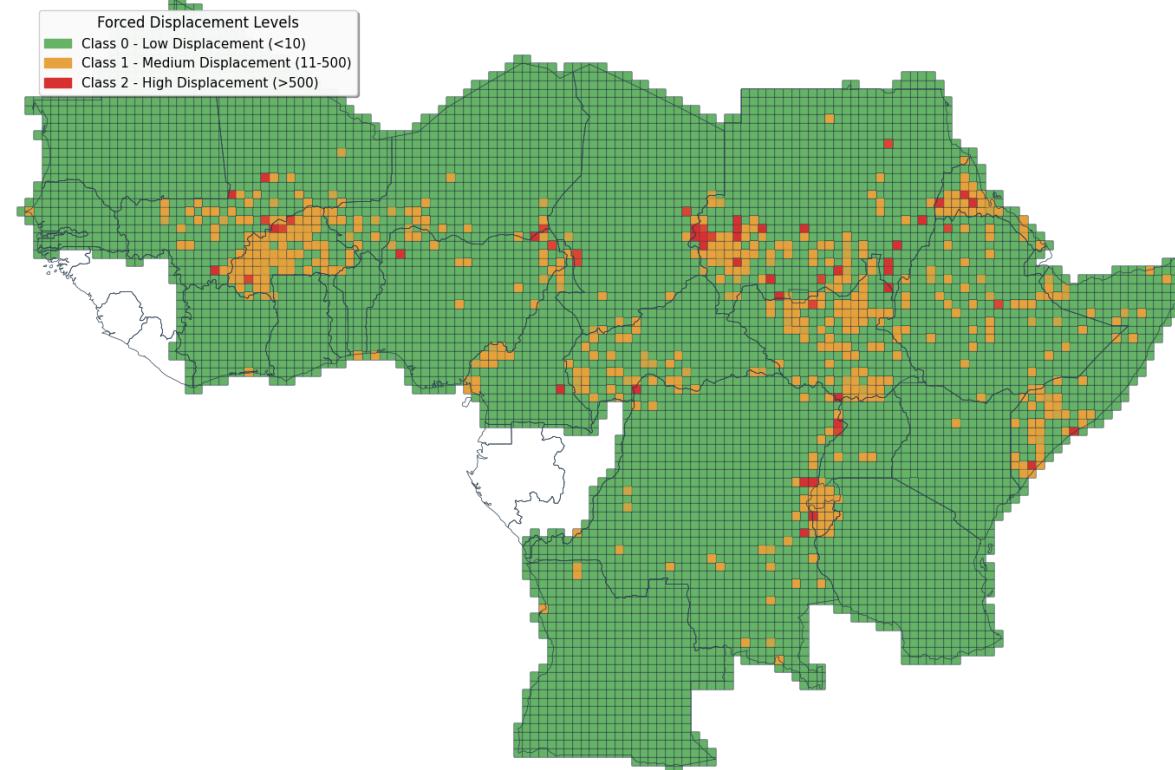
Max. temp. anomaly for last 12 months from 1980-90 temp. baseline



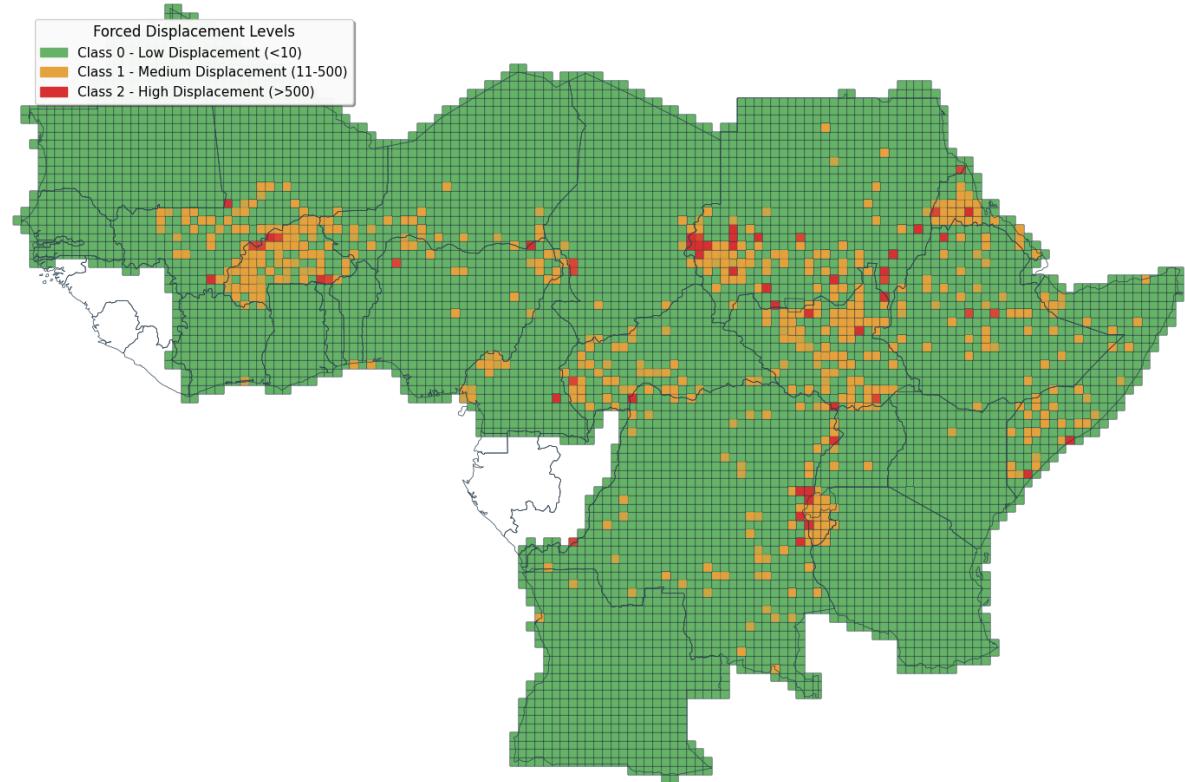
# Model Evaluation

Class	Model	1 Month	3 Months	6 Months
0	Baseline	0.9952	0.9953	0.9956
	LightGBM	<b>0.9997</b>	<b>0.9997</b>	<b>0.9996</b>
	Simple convLSTM	0.9992	0.9993	0.9990
	CNN	0.9993	0.9991	0.9979
	Hierarch. convLSTM	0.9986	0.9980	0.9990
	Ensemble	<u>0.9988</u>	<u>0.9996</u>	<u>0.9995</u>
1	Baseline	0.8155	0.7838	0.7600
	LightGBM	<b>0.9138</b>	<b>0.8773</b>	<b>0.8386</b>
	Simple convLSTM	0.8561	0.7983	0.6918
	CNN	0.8292	0.7560	0.6682
	Hierarch. convLSTM	0.8318	0.7719	0.7241
	Ensemble	<u>0.8623</u>	<u>0.8516</u>	<u>0.8110</u>
2	Baseline	0.8049	0.7193	0.6547
	LightGBM	<b>0.8741</b>	<u>0.7519</u>	<u>0.6588</u>
	Simple convLSTM	0.7990	0.6807	0.4819
	CNN	0.8167	0.6691	0.5630
	Hierarch. convLSTM	0.7625	0.6811	0.5839
	Ensemble	<u>0.8240</u>	<b>0.7570</b>	<b>0.6841</b>

# Model Evaluation



*Predicted displacement levels for 6 months ahead - Jun 2025*

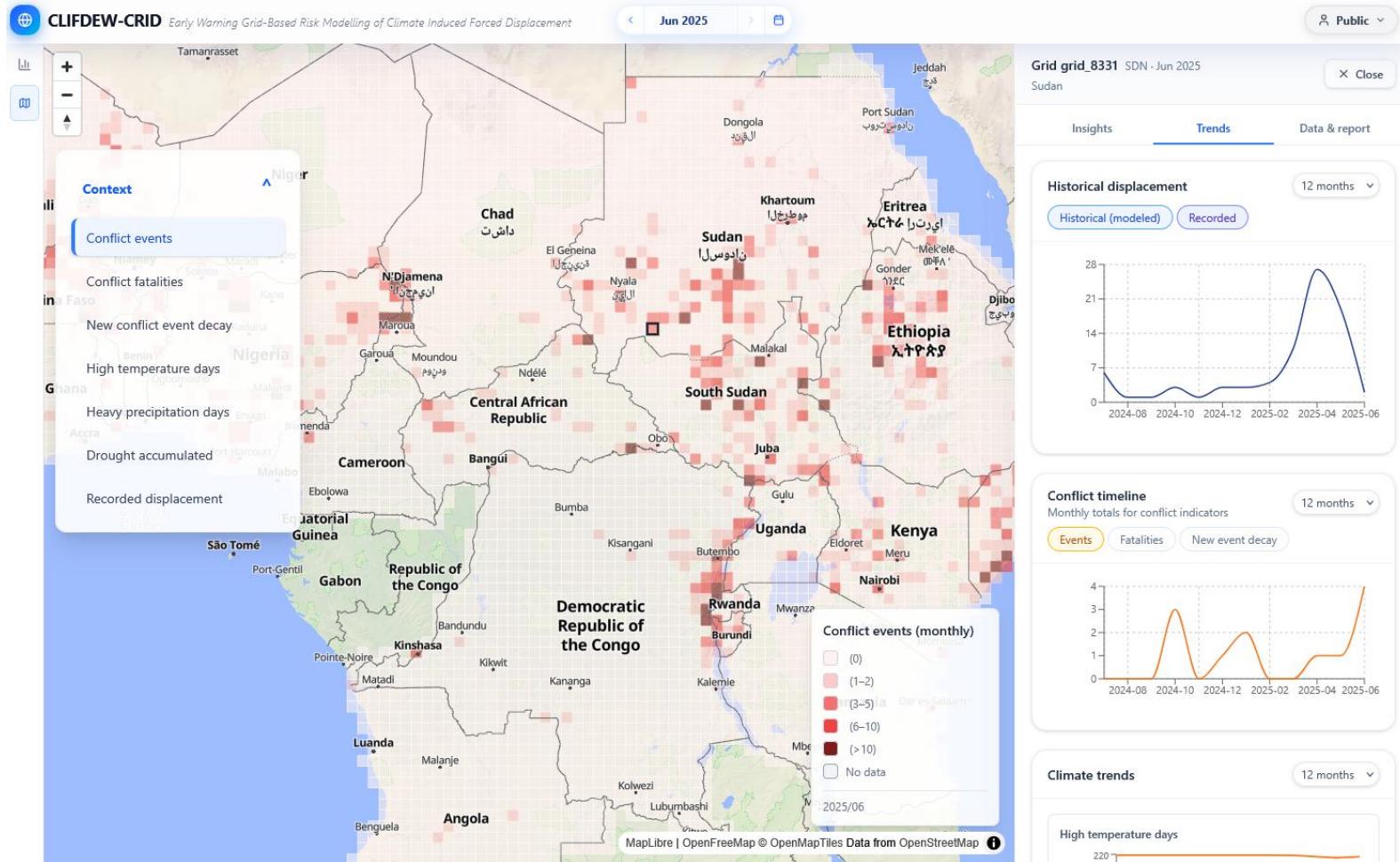


*Actual displacement levels - Jun 2025*

# Product

# Product

- Dashboard





# Global Early Warning System

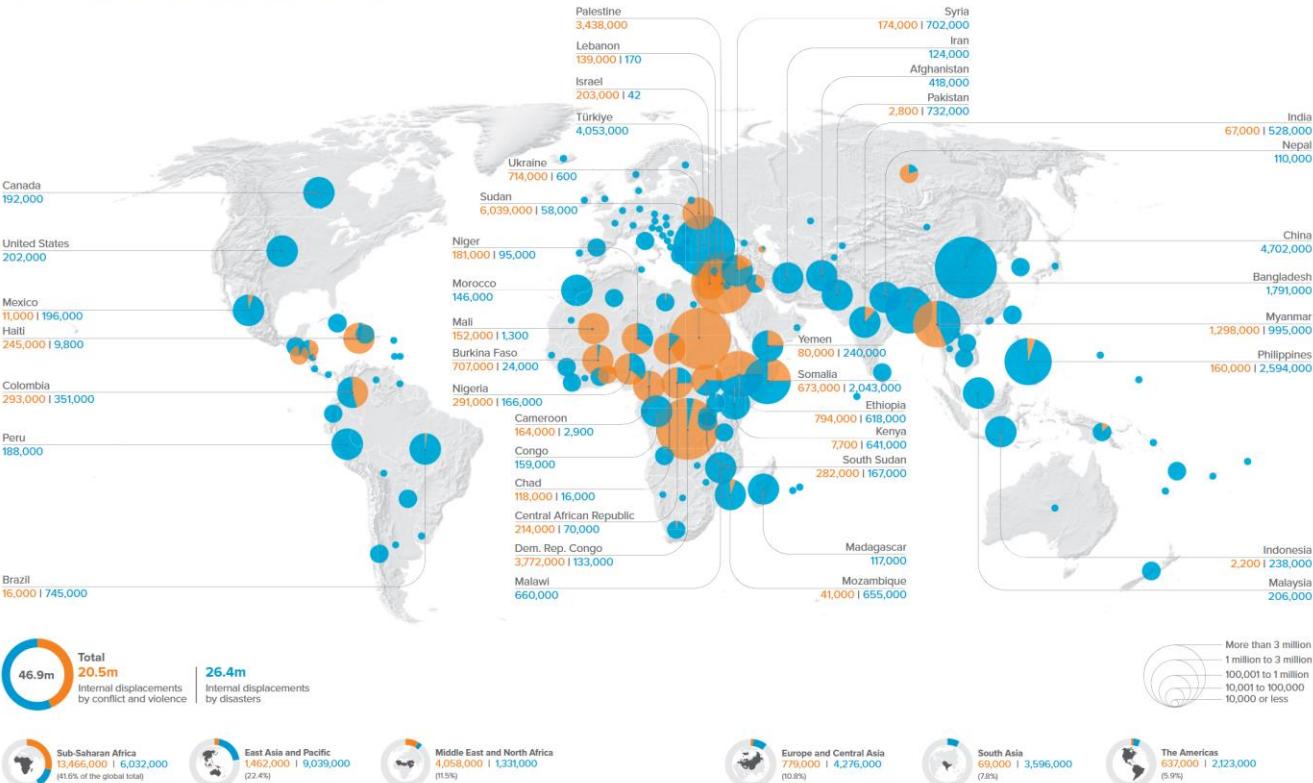
**Strengthening preparedness and response** in complex humanitarian emergencies.



# Background

- Conflict and violence triggered 20.5 million new internal displacements, across 45 countries.
- Disasters triggered 26.4 million new intern displacements across 148 countries.
- EWERS is funded by the Ministry of Foreign Affairs in Luxembourg.
- Collaboration between LIST and UNHCR.

## Internal displacements by conflict and disasters in 2023



# Goal | Impact - Displacement forecasting

We aim to predict  
**where/when/magnitude** of  
potential displacement 1 –  
2 weeks in advance

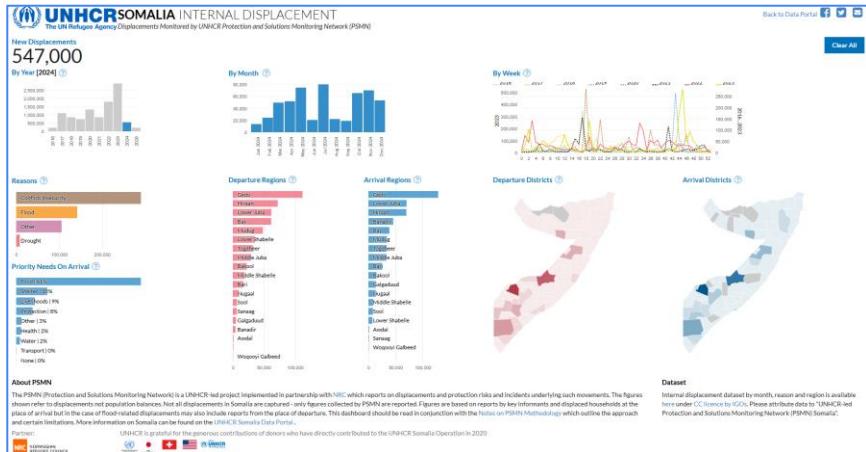


Displacement induced by  
**floods** (2025), conflict, and  
tropical cyclone

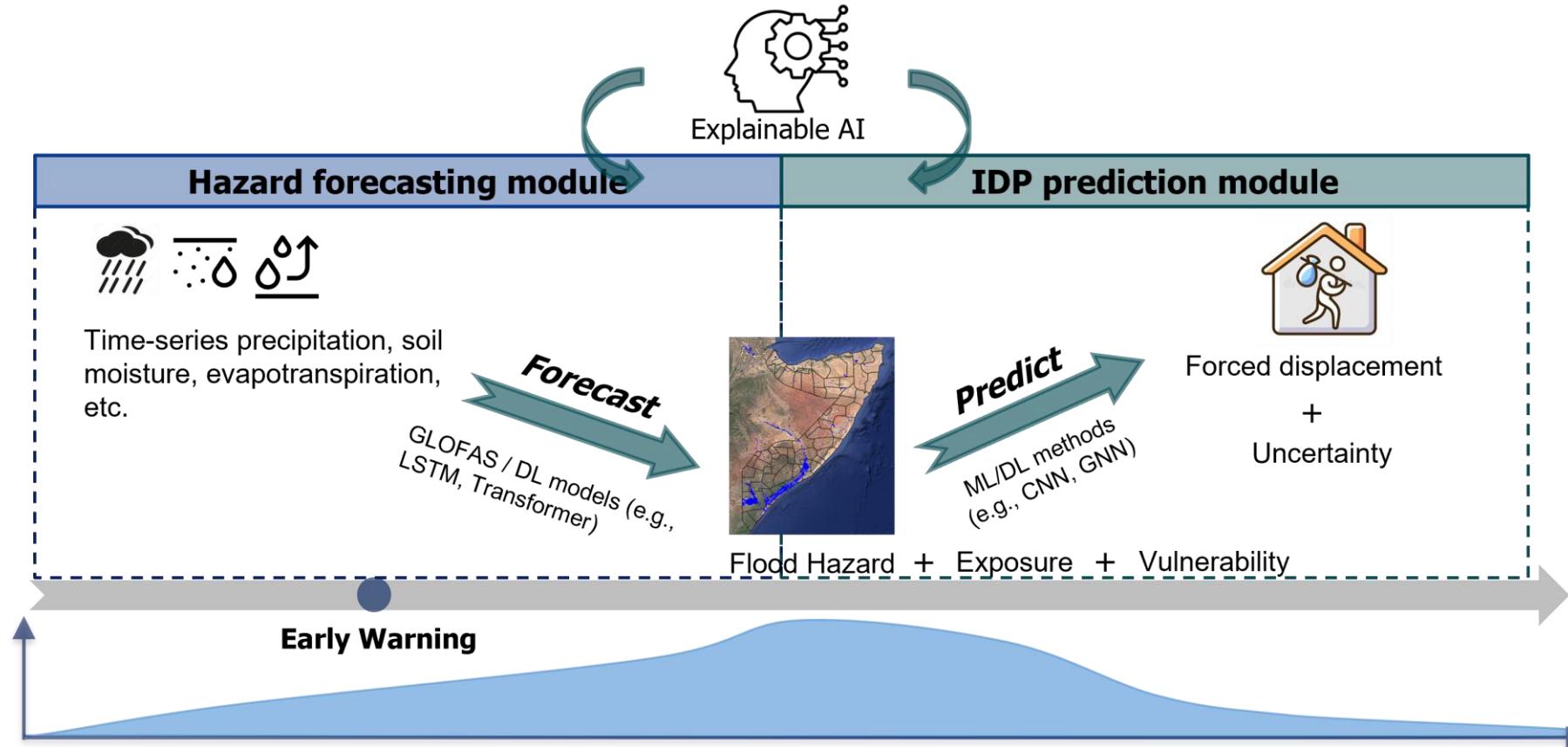
- ✓ Provide early warning
- ✓ Targeted preparedness actions
- ✓ Minimize response times
- ✓ Optimize supply process
- ✓ Avoid the duplication of humanitarian efforts
- ✓ Result in more life saving
- ✓ Achieve resource saving

# Prototype in Somalia

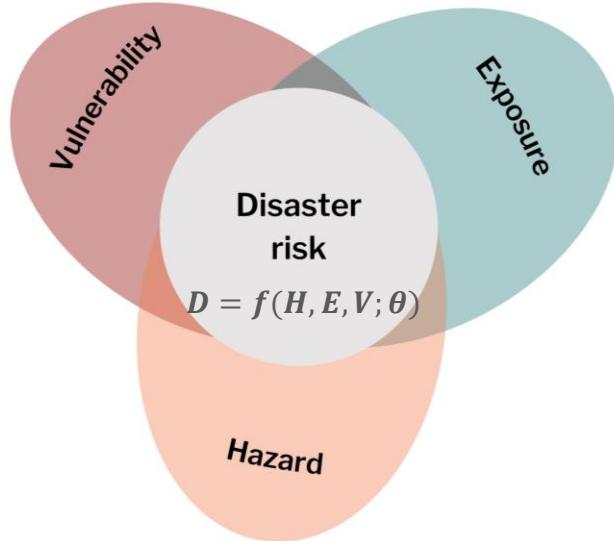
- UNHCR places Somalia **seventh** globally in terms of **displaced populations**
- Existing **internal displacement data** with solid support from country-based colleagues
- Cyclic flood causing forced internal **displacement**
- Challenge to address **complex overlapping drivers** of displacement



# Impact-based Forecasting & Anticipatory Action



# Impact-based Forecasting & Anticipatory Action



$$D = f(H, E, V; \theta)$$

**D:** IDP to be predicted

**H:** Hazard intensity (e.g., flood depth, precipitation)

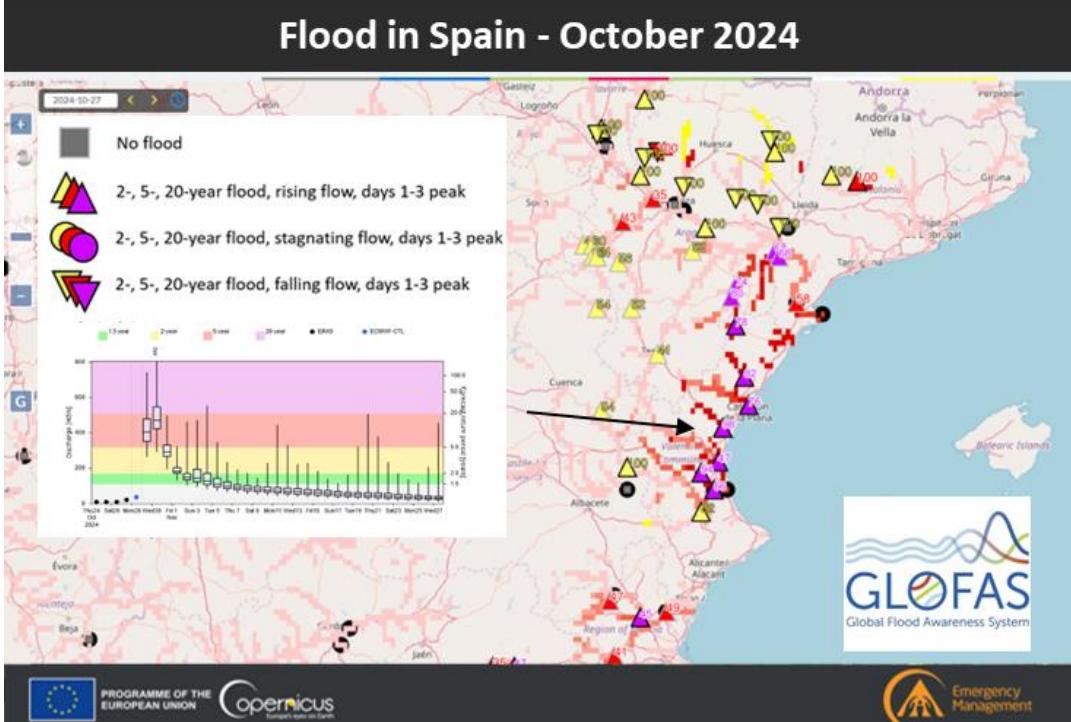
**E:** Exposure (e.g., population, building)

**V:** Vulnerability (e.g., socioeconomic and environmental factors)

$f(\theta)$ : Impact function (e.g., neural networks with parameters  $\theta$ )

# Flood monitoring & forecasting

## Flood in Spain - October 2024



Flood prediction model

Sentinel-1 Derived Map of Flood Affected Urban Area - Valencia, Spain



Satellite-based observation of flood extent

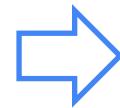
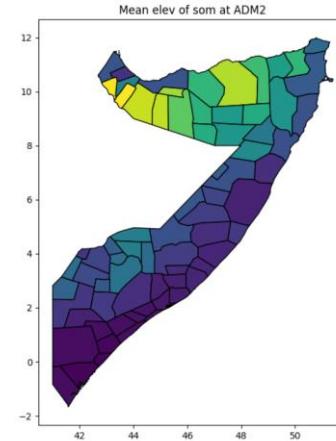
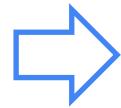
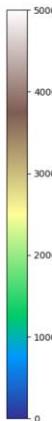
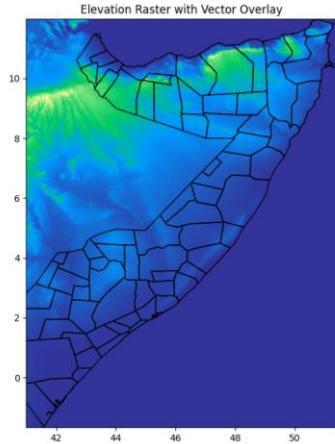
# Data Preparation

	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	
1	Month	End	Year	Yr	Week	Current	Current	Previous	Previous	Reason	Priority	Discharge	Water depth	Precipita	Populatric	Building Fraction	AWI	Agricultural land	Total Individuals
1489	30/04/2016	2016	201616	Awdal	Borama	Awdal	Borama	Flood	Food									5	
1684	31/05/2016	2016	201618	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									135	
1685	31/05/2016	2016	201618	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									144	
1694	31/05/2016	2016	201619	Lower Juba:Jamaame	Lower Juba:Jamaame	Lower Juba:Jamaame	Lower Juba:Jamaame	Flood	Shelter									133	
1697	31/05/2016	2016	201618	Middle Sha Jowhar	Middle Sha Jowhar	Middle Sha Jowhar	Middle Sha Jowhar	Flood	Other									810	
1700	31/05/2016	2016	201619	Bakool	Ceel Barde Bakool	Ceel Barde Bakool	Ceel Barde Bakool	Flood	Food									150	
1711	31/05/2016	2016	201619	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									6	
1712	31/05/2016	2016	201619	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									5	
1713	31/05/2016	2016	201619	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									4	
1734	31/05/2016	2016	201619	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									119	
1757	31/05/2016	2016	201619	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									495	
1758	31/05/2016	2016	201619	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									227	
1841	31/05/2016	2016	201620	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									5	
1842	31/05/2016	2016	201620	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									5	
1843	31/05/2016	2016	201620	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									5	
1860	31/05/2016	2016	201620	Mudug	Gaalkacyo	Hiraan	Belet Weyn	Flood	Livelihoods									6	
1914	31/05/2016	2016	201620	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									325	
1962	31/05/2016	2016	201621	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									7	
1963	31/05/2016	2016	201621	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									4	
1964	31/05/2016	2016	201621	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									5	
1965	31/05/2016	2016	201621	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									6	
1966	31/05/2016	2016	201621	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									5	
1968	31/05/2016	2016	201621	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									8	
1980	31/05/2016	2016	201621	Mudug	Gaalkacyo	Hiraan	Belet Weyn	Flood	Other									8	
1988	31/05/2016	2016	201621	Bakool	Tayeeqglow	Hiraan	Belet Weyn	Flood	Food									183	
1999	31/05/2016	2016	201621	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									5	
2000	31/05/2016	2016	201621	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Middle Juba:Saakow	Flood	Food									4	

- Weekly IPD data:** UNHCR-PSMN project
- Hazard information:** river discharge from GLOFAS, precipitation from ECMWF/ERA5
- Exposure:** population density, built-up area, elevation, crop land
- Vulnerability:** poverty (e.g., HDI, IWI, GINI), conflict situation

# Data Preparation

## Extraction of topography at Admin-2 level

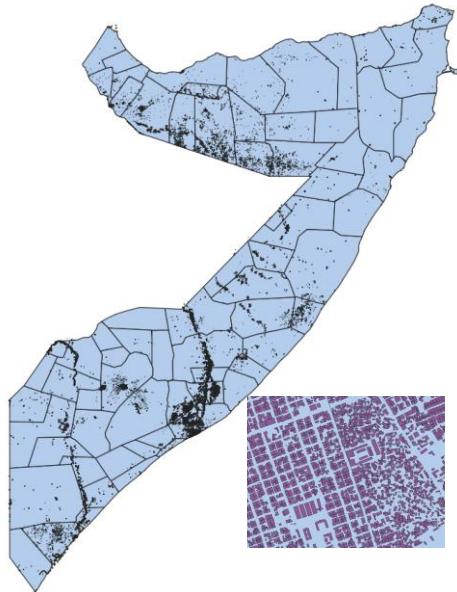


	A	B	C
1	ADM2_EN	ADM2_PCODE	mean_DEM
2	Baki	SO1102	536.211
3	Borama	SO1101	1213.777
4	Lughaye	SO1103	181.838
5	Zeylac	SO1104	332.345
6	Ceel Barde	SO2502	525.27
7	Rab Dhuure	SO2505	370.443
8	Tayeeglow	SO2503	426.706
9	Waajid	SO2504	444.333
10	Xudur	SO2501	527.783
11	Bondhere	SO2201	30.438
12	Cabdulasis	SO2202	4.438
13	Daynile	SO2203	100.999
14	Dharkenley	SO2204	45.752
15	Hamar Jabjab	SO2205	17.438
16	Hamar Weyne	SO2206	8.875
17	Hawl Wadaag	SO2207	51.062
18	Heliwa	SO2208	39.514
19	Hodan	SO2209	66.062

Topography—through variations in elevation —shapes how floodwaters spread across a landscape, affecting which areas are inundated, how severely populations are impacted, and where displacement is most likely to occur.

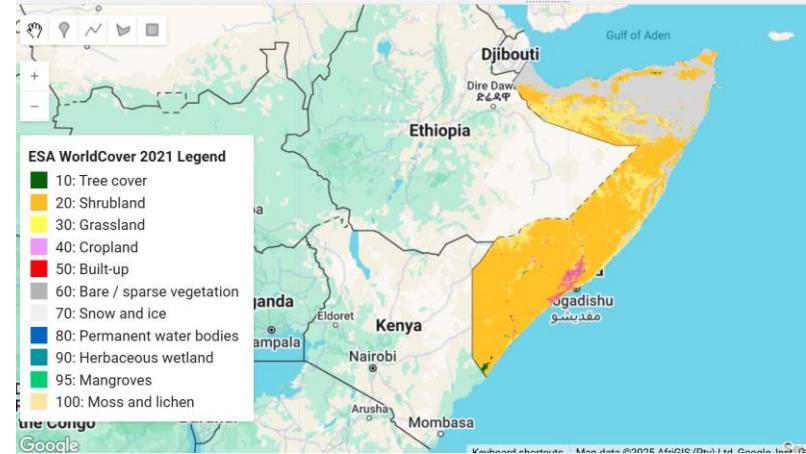
# Data Preparation

Building footprint



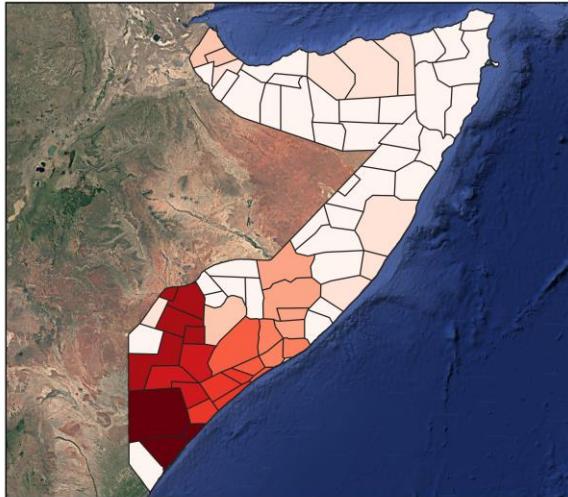
Sources : OpenStreetMap / ESA WorldCover

Landcover

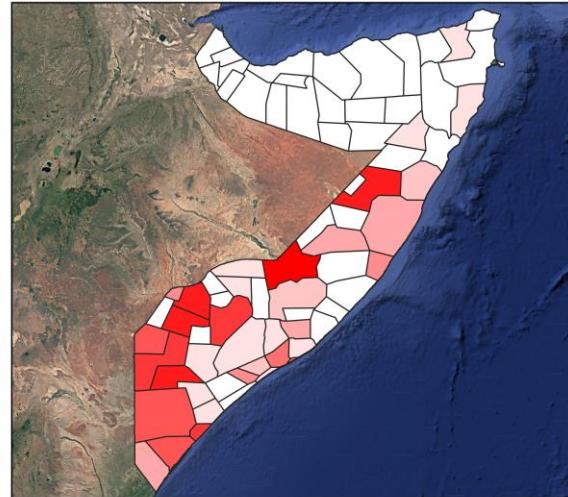


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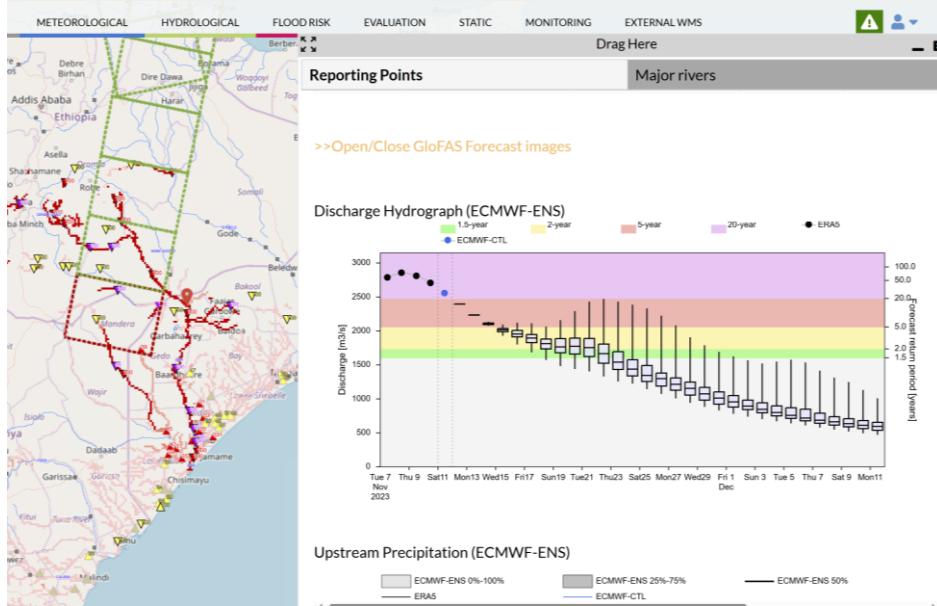
GLOFAS discharge



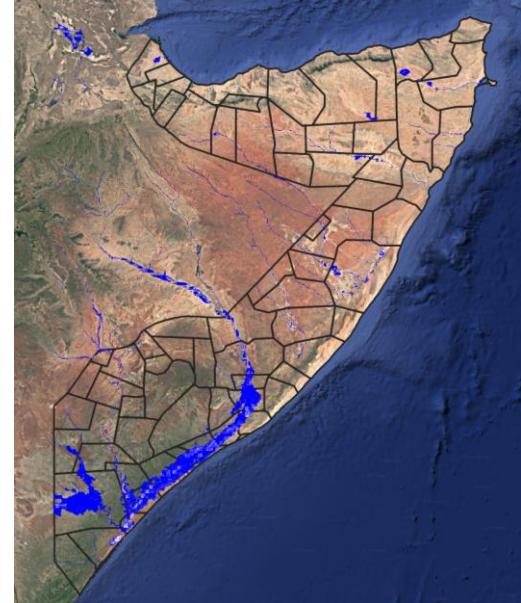
Flood-induced IDP in November 2023 from  
UNHCR-PSMN



# Data Preparation



GLOFAS discharge forecast

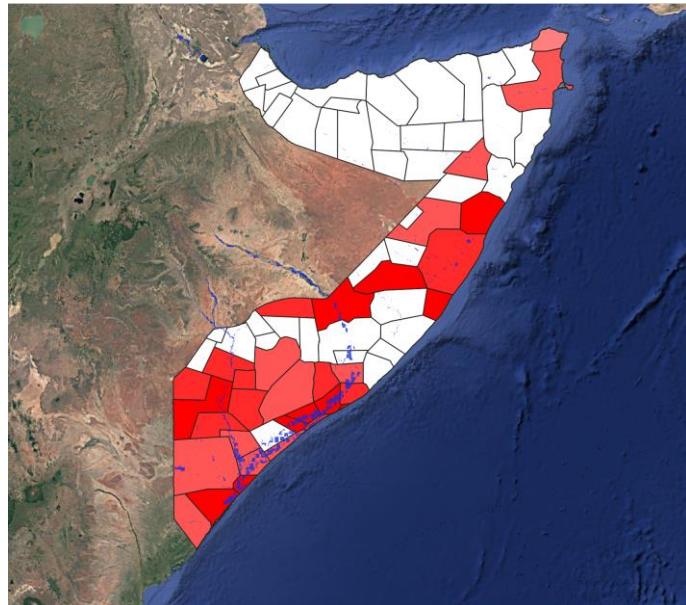


Flood Water Depth

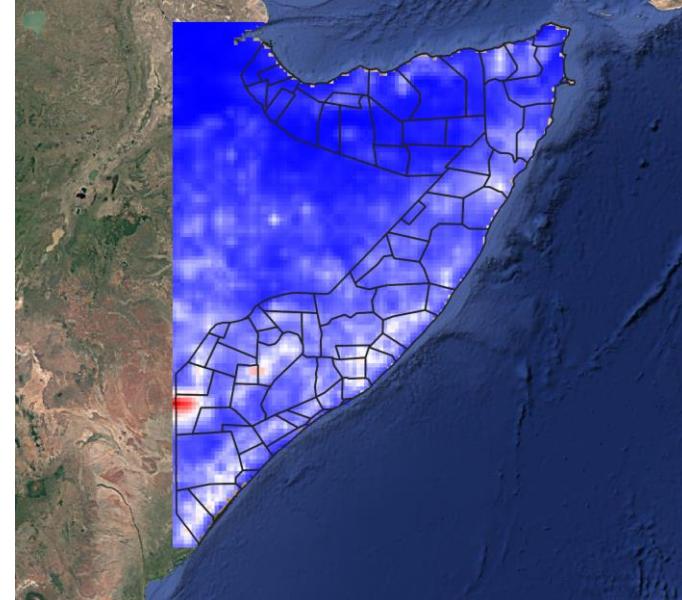
# Data Preparation

Week 2023-36, starts at 12/11/2023

Water depth  $\geq 10$  cm on top of IDP records



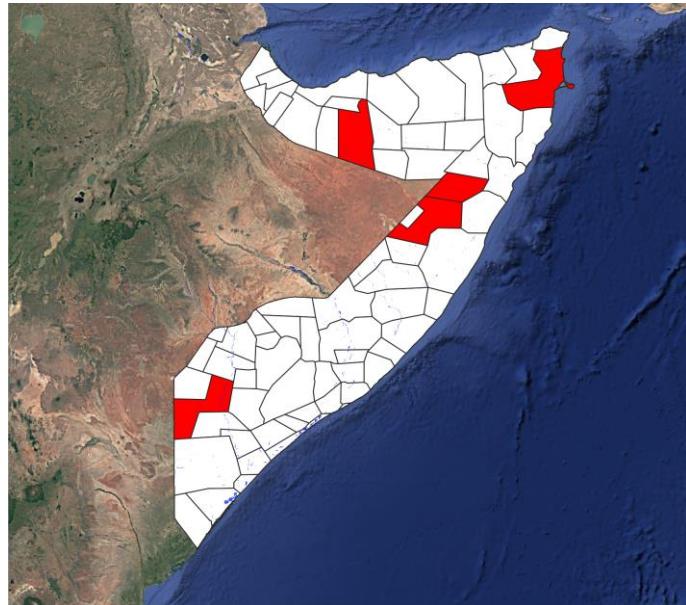
Maximum precipitation



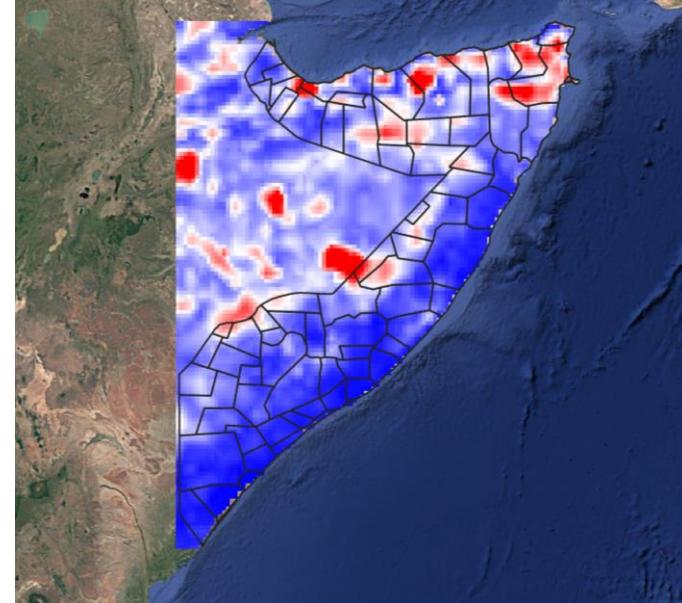
# Data Preparation

Week 2023-12, starts at 19/03/2023

Water depth  $\geq 10$  cm on top of IDP records



Maximum precipitation



# Data Preparation

FLOOD	EXPOSURE	VULNERABILITY
Environmental and climatic triggers of displacement	Population and assets located in hazard zones	Ability to prepare for, respond to, and recover from displacement
<ul style="list-style-type: none"> <li>● <b>Riverine flood – Flood fraction</b> (GloFas)</li> <li>● <b>Precipitation</b> (ERA5)</li> <li>● <b>Wind speed</b> (ERA5)</li> <li>● <b>Soil Moisture</b> (ERA5)</li> <li>● <b>Convective available potential energy (CAPE)</b> (ERA5)</li> <li>● <b>Air Temperature</b> (ERA5)</li> <li>● <b>Evaporation</b> (ERA5)</li> </ul>	<ul style="list-style-type: none"> <li>● <b>Demographics</b> (Age groups proportion -WorldPop)</li> <li>● <b>Building density</b> (ESA WorldCover)</li> <li>● <b>Populated area</b> (HDX)</li> <li>● <b>Topography</b> (Elevation — Earthenv)</li> <li>● <b>Travel time to the nearest market</b> (HDX)</li> <li>● <b>Week Number</b></li> </ul>	<ul style="list-style-type: none"> <li>● <b>Poverty and income</b> (HDI, IWI— World Bank, UNDP)</li> <li>● <b>Conflict exposure</b> (conflict fatalities - ACLED)</li> </ul>

● Static or Annual Updates

● Weekly or Frequent Updates - cannot be forecasted but can be integrated into the model actively.

● Forecasted Variables - can be used to predict displacement before it happens.

# Modeling Experiments

- **IDP data** (PSMN, 2016-2025): data between 2019-2024 (1109 weekly records) is used for experiments (GLOFAS flood forecast is only available from 2019).
- **Target:** One week lead time IDP forecast.
- **Experiments:** Bootstrapping with 100 resampling for hypothesis test and model comparison, with 80/20 data random split of training and test dataset for each running.

# Modeling Experiments

## State of the art

nature communications



Article

<https://doi.org/10.1038/s41467-023-43809-8>

### Exploring interactions between socioeconomic context and natural hazards on human population displacement

Received: 6 October 2022

Accepted: 20 November 2023

Published online: 04 December 2023

Michele Ronco , José María Tárraga , Jordi Muñoz , María Piles , Eva Sevillano Marco , Qiang Wang , María Teresa Miranda Espinosa , Sylvain Ponserre & Gustau Camps-Valls

nature communications



Article

<https://doi.org/10.1038/s41467-025-64015-8>

### Socioeconomic predictors of vulnerability to flood-induced displacement

Received: 8 October 2024

Accepted: 4 September 2025

Published online: 16 September 2025

Benedikt Mester , Katja Frieler , Oliver Korup , Bina Desai & Jacob Schewe

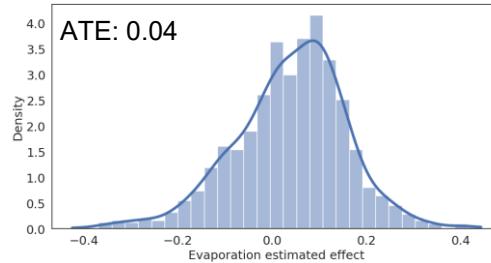
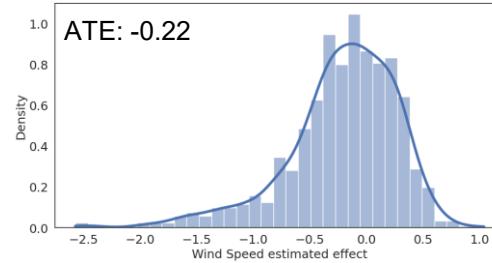
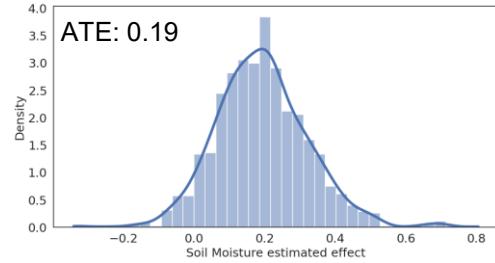
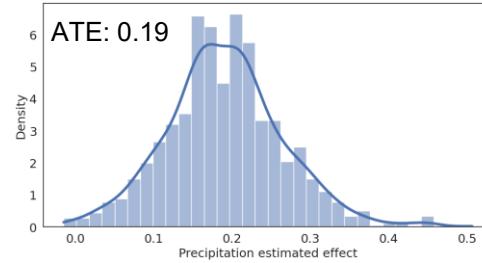
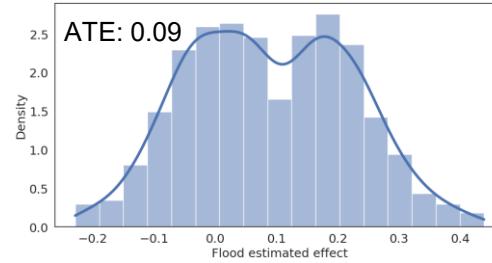
- Based on real observations
- Admin 1 to national level, global scale
- Random Forest

$R^2$ : ~ 0.31

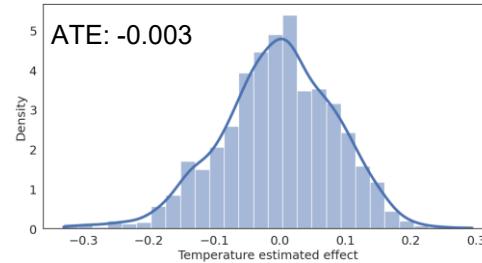
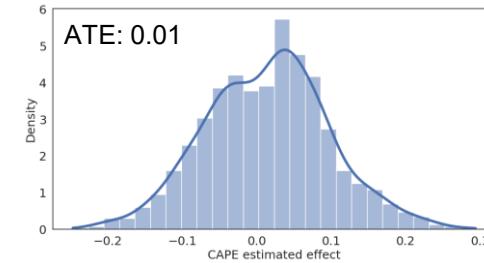
$R^2$ : 0.28-0.34

# Modeling Experiments

Causal effect estimation of weather-related variables through double machine learning



ATE: Average Treatment Effect



# Modeling Experiments

FLOOD	EXPOSURE	VULNERABILITY
Environmental and climatic triggers of displacement	Population and assets located in hazard zones	Ability to prepare for, respond to, and recover from displacement
<ul style="list-style-type: none"> <li>● <b>Riverine flood – Flood fraction</b> (GloFas)</li> <li>● <b>Precipitation</b> (ERA5)</li> <li>● <b>Wind speed</b> (ERA5)</li> <li>● <b>Soil Moisture</b> (ERA5)</li> <li>● <b>Convective available potential energy (CAPE)</b> (ERA5)</li> <li>● <b>Air Temperature</b> (ERA5)</li> <li>● <b>Evaporation</b> (ERA5)</li> </ul>	<ul style="list-style-type: none"> <li>● <b>Demographics</b> (Age groups proportion -WorldPop)</li> <li>● <b>Building density</b> (ESA WorldCover)</li> <li>● <b>Populated area</b> (HDX)</li> <li>● <b>Topography</b> (Elevation — Earthenv)</li> <li>● <b>Travel time to the nearest market</b> (HDX)</li> <li>● <b>Week Number</b></li> </ul>	<ul style="list-style-type: none"> <li>● <b>Poverty and income</b> (HDI, IWI— World Bank, UNDP)</li> <li>● <b>Conflict exposure</b> (conflict fatalities - ACLED)</li> </ul>

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● Weekly or Frequent Updates - cannot be forecasted but can be integrated into the model actively.

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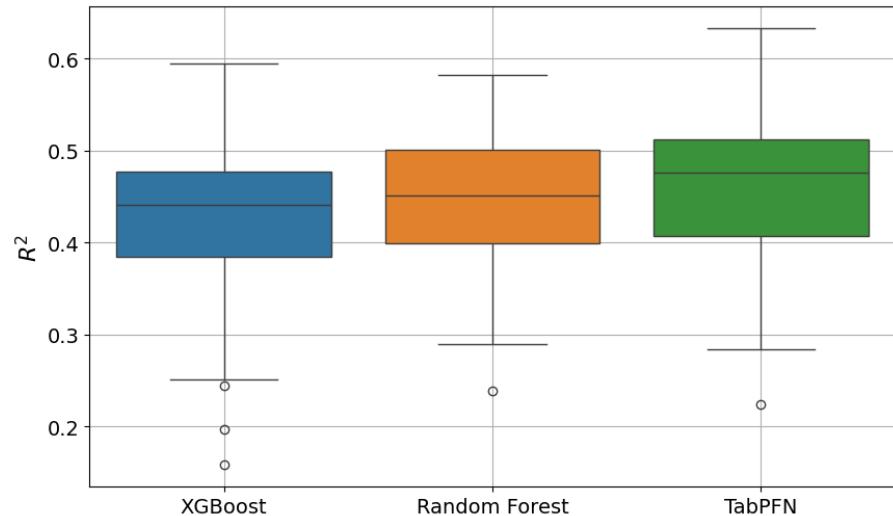
# Modeling Experiments

## Regression models comparison

$R^2: 0.42 \pm 0.08$

$0.44 \pm 0.07$

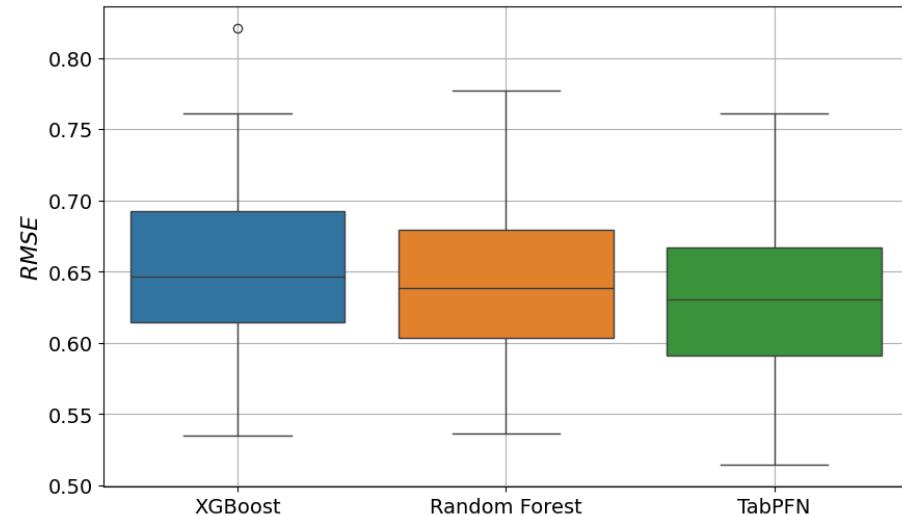
$0.46 \pm 0.08$



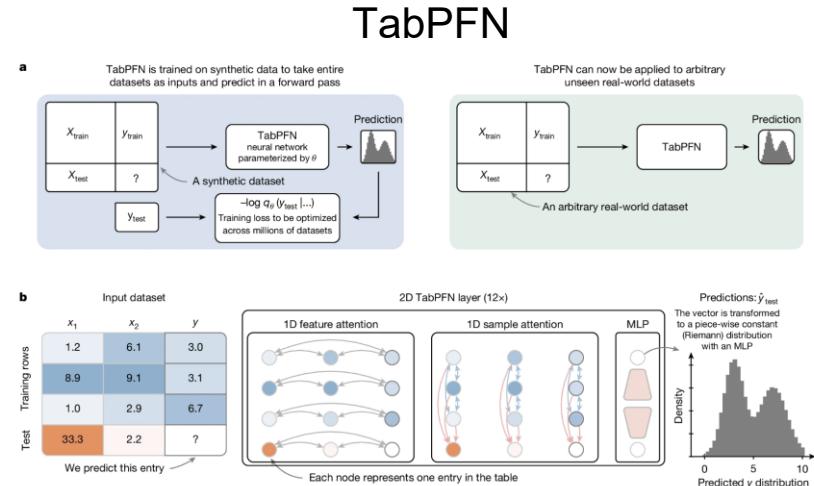
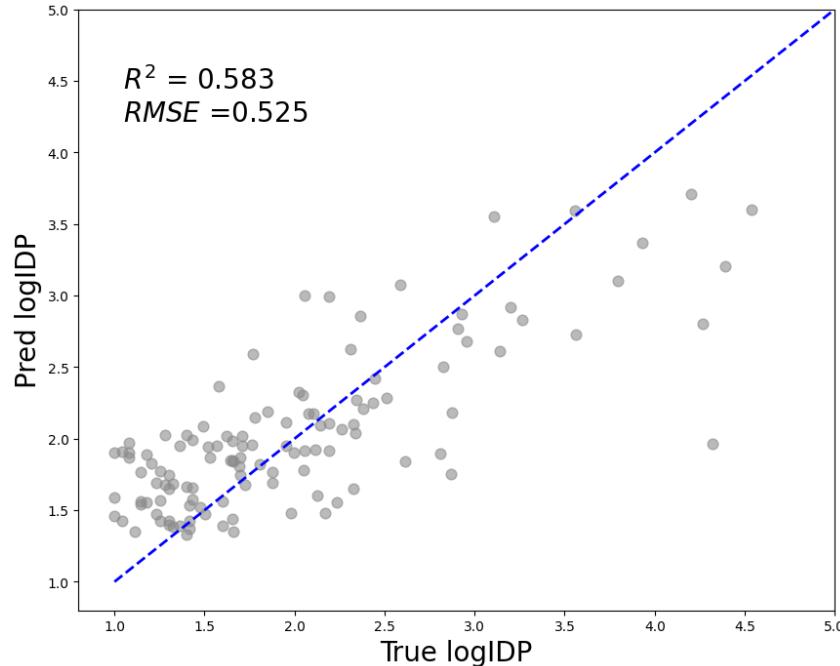
$RMSE: 0.65 \pm 0.06$

$0.64 \pm 0.05$

$0.62 \pm 0.07$



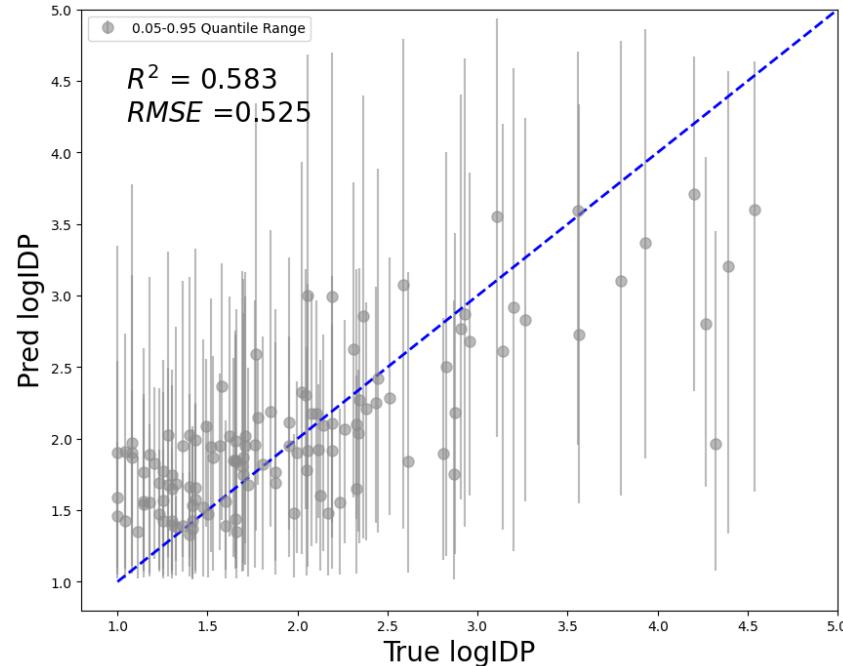
# Modeling Experiments



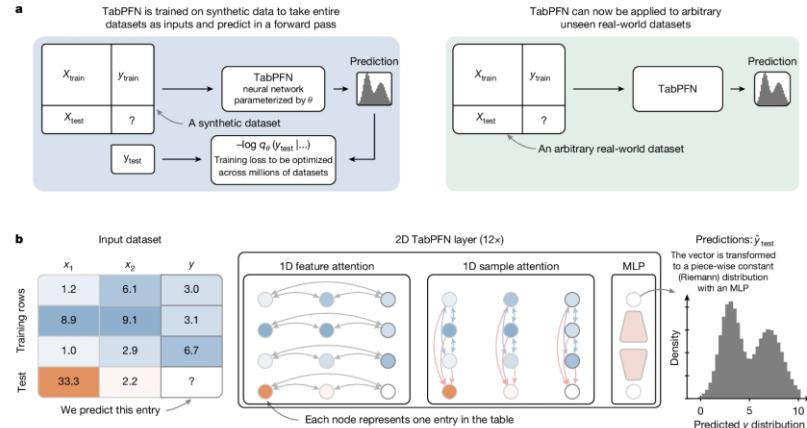
- Pretrained tabular foundation model using synthetical data generated through causal models
- In-context learning

- Over-estimation in small events (e.g., IDP < 100), under-estimation in large events (e.g., IDP > 5000)

# Modeling Experiments



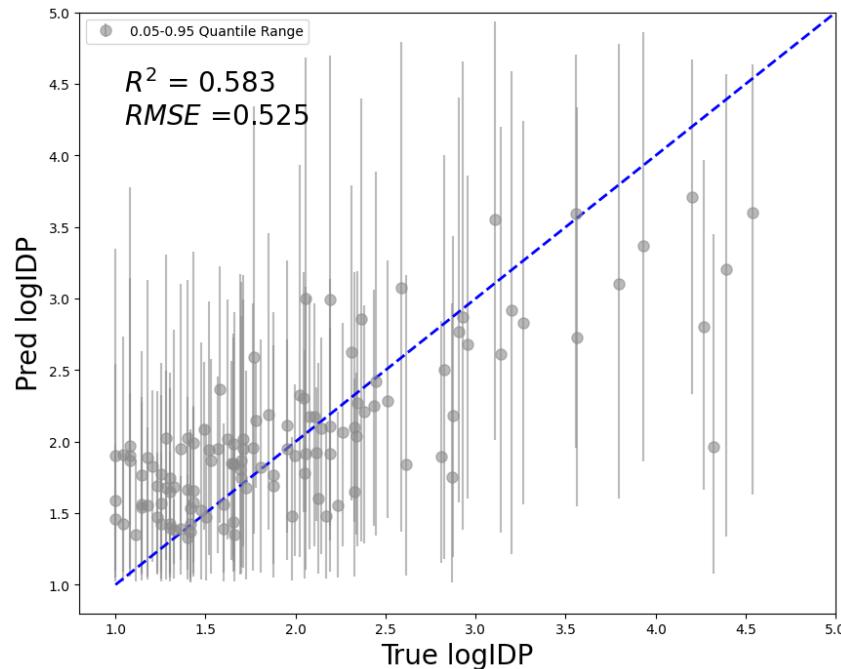
## TabPFN



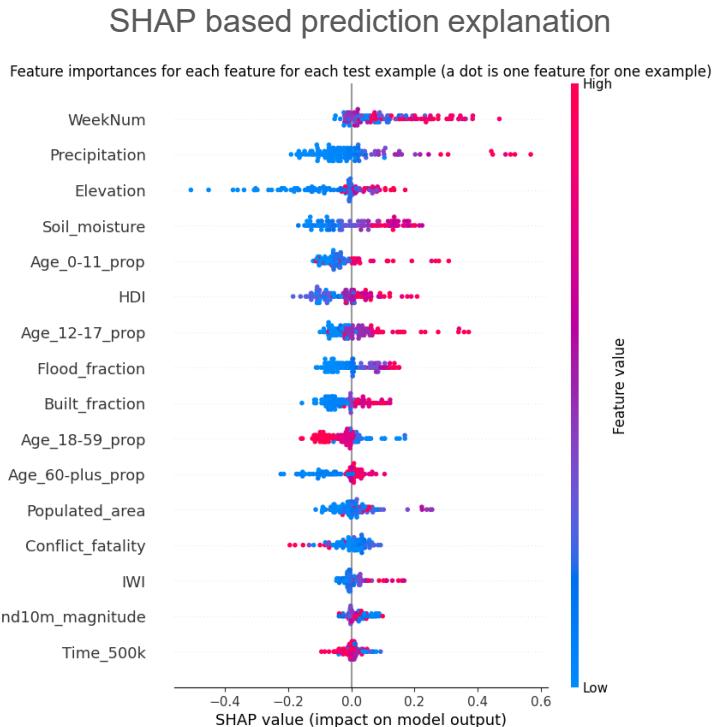
- Pretrained tabular foundation model using synthetical data generated through causal models
- In-context learning

- Over-estimation in small events (e.g., IDP < 100), under-estimation in large events (e.g., IDP > 5000)
- Apply 5% and 95% quantile regression to offer 90% prediction interval

# Modeling Experiments

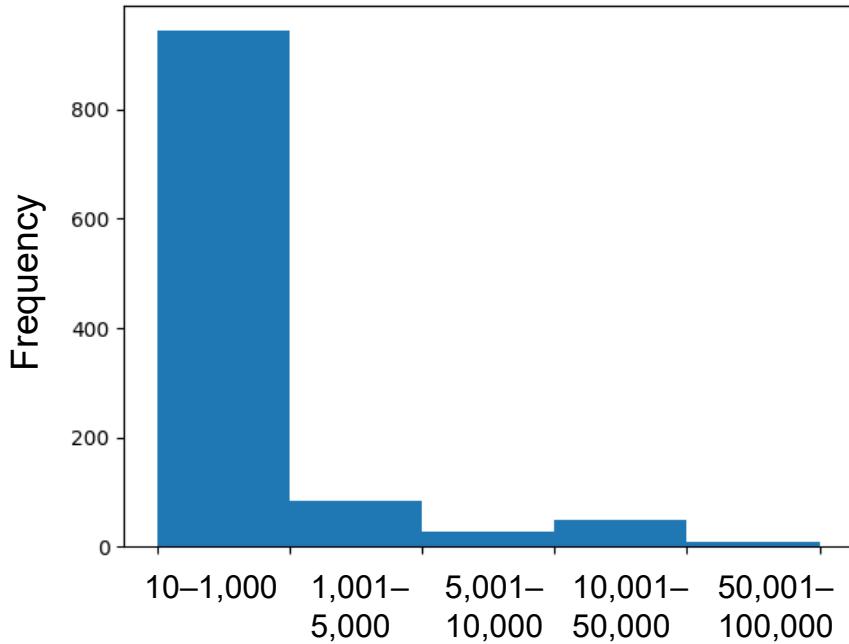


- Over-estimation in small events (e.g., IDP < 100), under-estimation in large events (e.g., IDP > 5000)
- Apply 5% and 95% quantile regression to offer 90% prediction interval



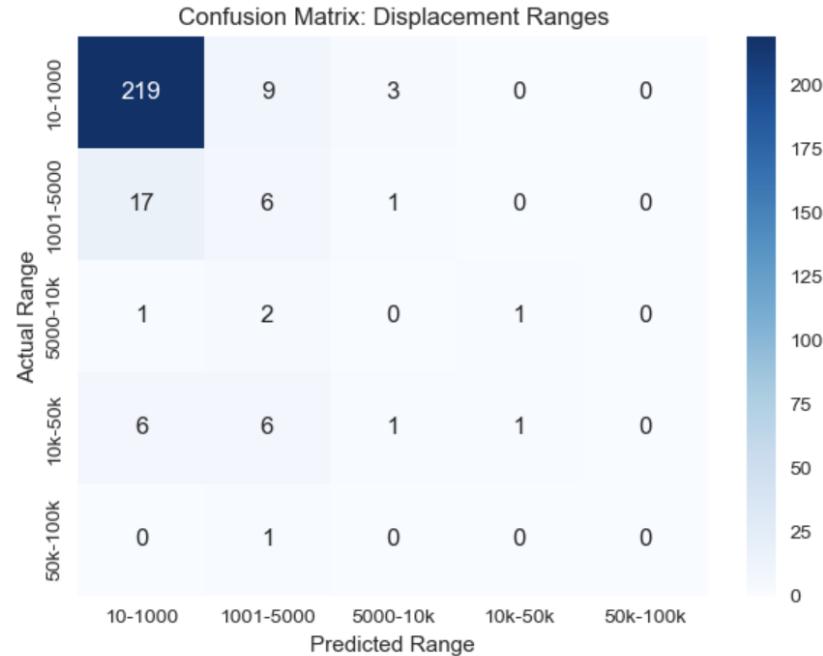
# Modeling Experiments

- Categorize IDP into ranges based on several operational thresholds
- Dataset is severely skewed to small events (e.g., IDP < 1000)



# Modeling Experiments

- Evaluation on IDP ranges shows high accuracy in classifying small events with IDP < 1000.
- Performance drops dramatically in large events.
- Overall accuracy: 0.82 – dominated by small events.



# Roadmap

- Enhance IPD data collection: IDMC disaggregated data is supposed to be released in Feb 2026.
- Have a better understanding of root causes of IDP: close collaboration with UNHCR and local authorities.
- Integrate other important social-economic and environmental predictors that are currently ignored.
- Improve Flood forecast accuracy: GLOFAS v 5.0 will be in place in April 2026. Account for both riverine and flash flood: Google flash flood model will be released in early 2026.
- Test model generalization across different countries (e.g., Chad, CAR)
- Investigate IDP induced by other hazards as well as armed conflict.

# Thank you!



Asuka Imai, Senior Project Manager ([imai@unhcr.org](mailto:imai@unhcr.org))

Ines Krissaane Ph.D. Associate Data Scientist

With the support of

GDS UNHCR

Innovation UNHCR

**UNHCR Somalia PSMN Displacement  
and Protection Information**

SWALIM FAO

BRICS / Norwegian Refugee Council



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Aolin Jia, R&D Scientist

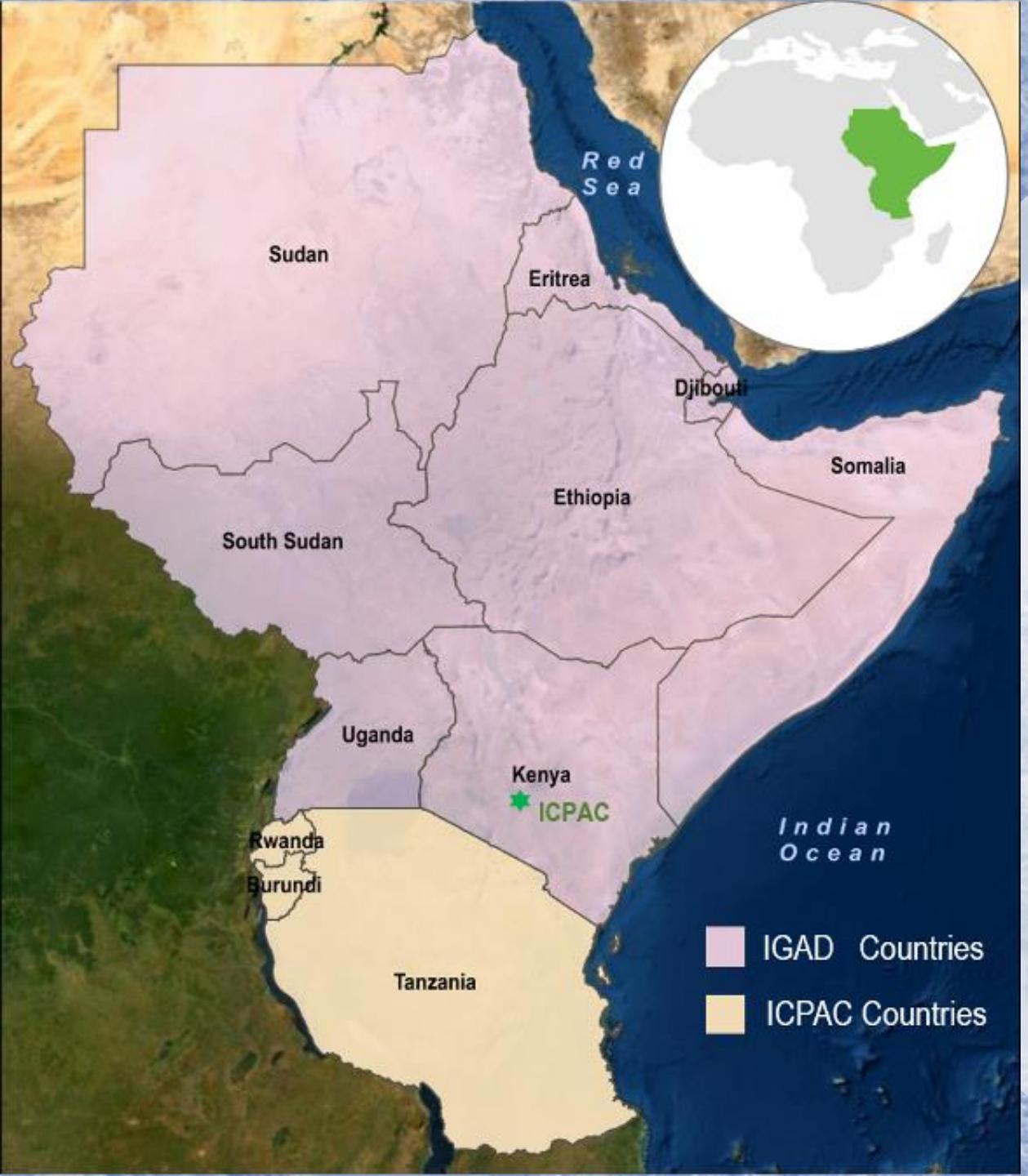




ICPAC



## IGAD CLIMATE PREDICTION AND APPLICATIONS CENTRE (ICPAC)



- Established in **1989** as DMC, in Nairobi.
- Adopted by IGAD in **2007** as IGAD Climate Prediction and Application Centre.



RCC in May 2017



Observer status

**AUDA - NEPAD**  
AFRICAN UNION DEVELOPMENT AGENCY

Central and East Africa - Centre of Excellence

# ICPAC units



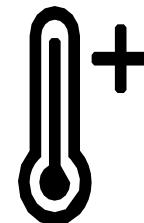
**Climate Diagnostic and Prediction**



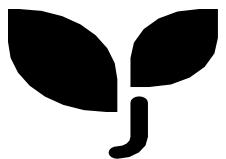
**Disaster Risk Management**



**Climate Applications**



**Climate Change**



**Data Management,  
Remote Sensing and  
Geospatial**



**The IGAD Food Security,  
Nutrition and resilience  
analysis Hub, IFRAH**



# ICPAC SERVICES



ICPAC



Climate  
Forecasting



Disaster Risk  
Management



Water Resources



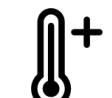
Agriculture and  
Food Security



Environmental  
Monitoring



Capacity Development



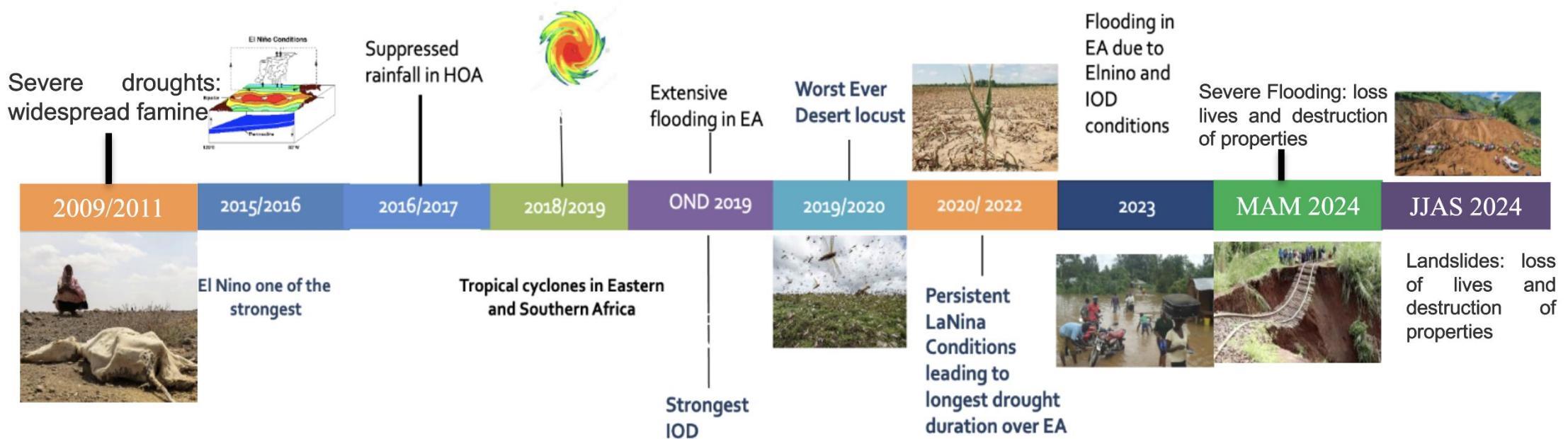
Climate Change



Climate Information  
Dissemination

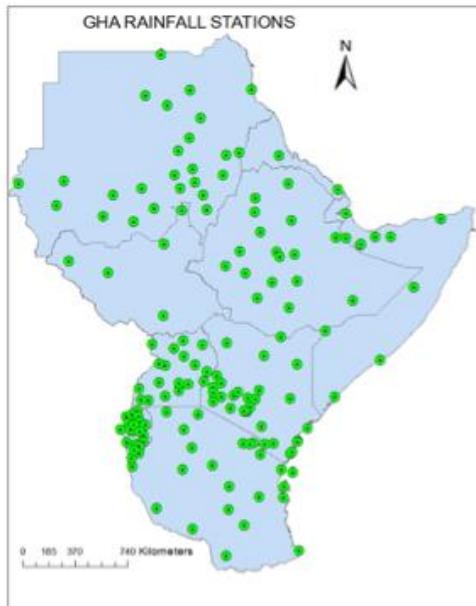
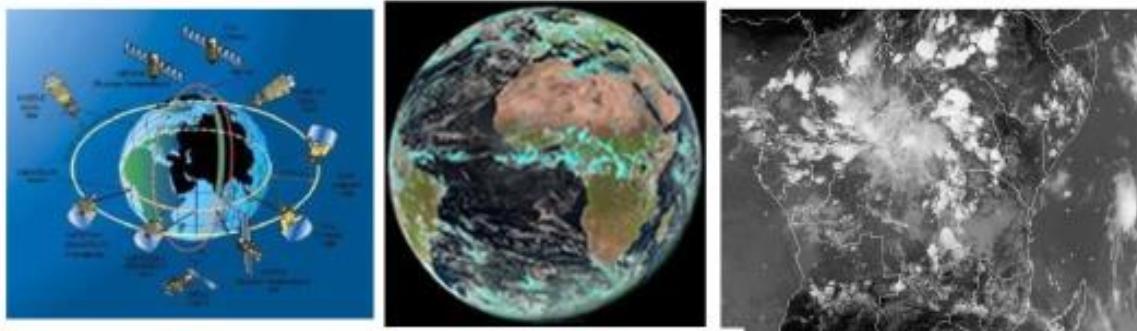
**MISSION:** Foster climate services and knowledge to enhance  
community resilience in the region

# RECENT INTERPLAY OF CLIMATE RISKS OVER IGAD REGION

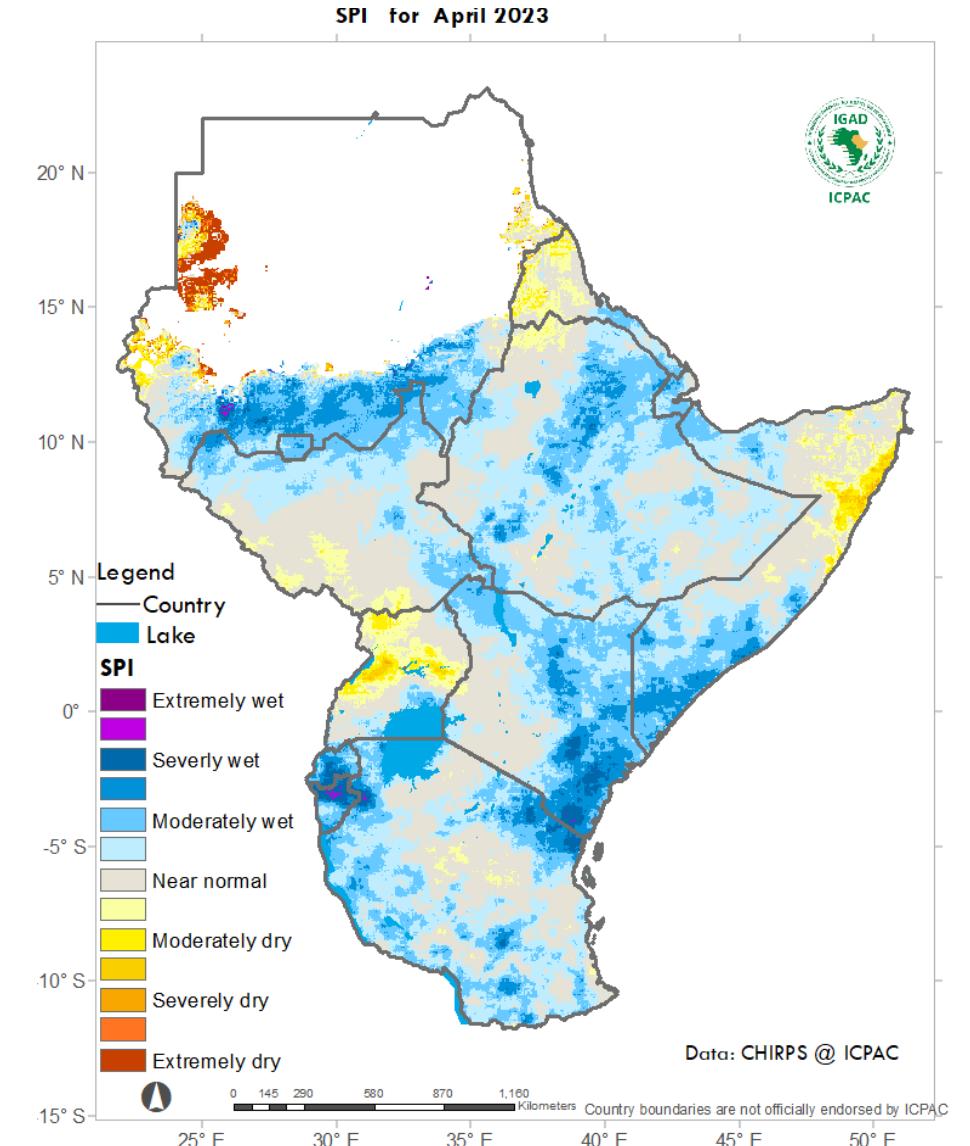


# DATA SERVICES: FROM STATION DATA TO GRIDDED DATASETS

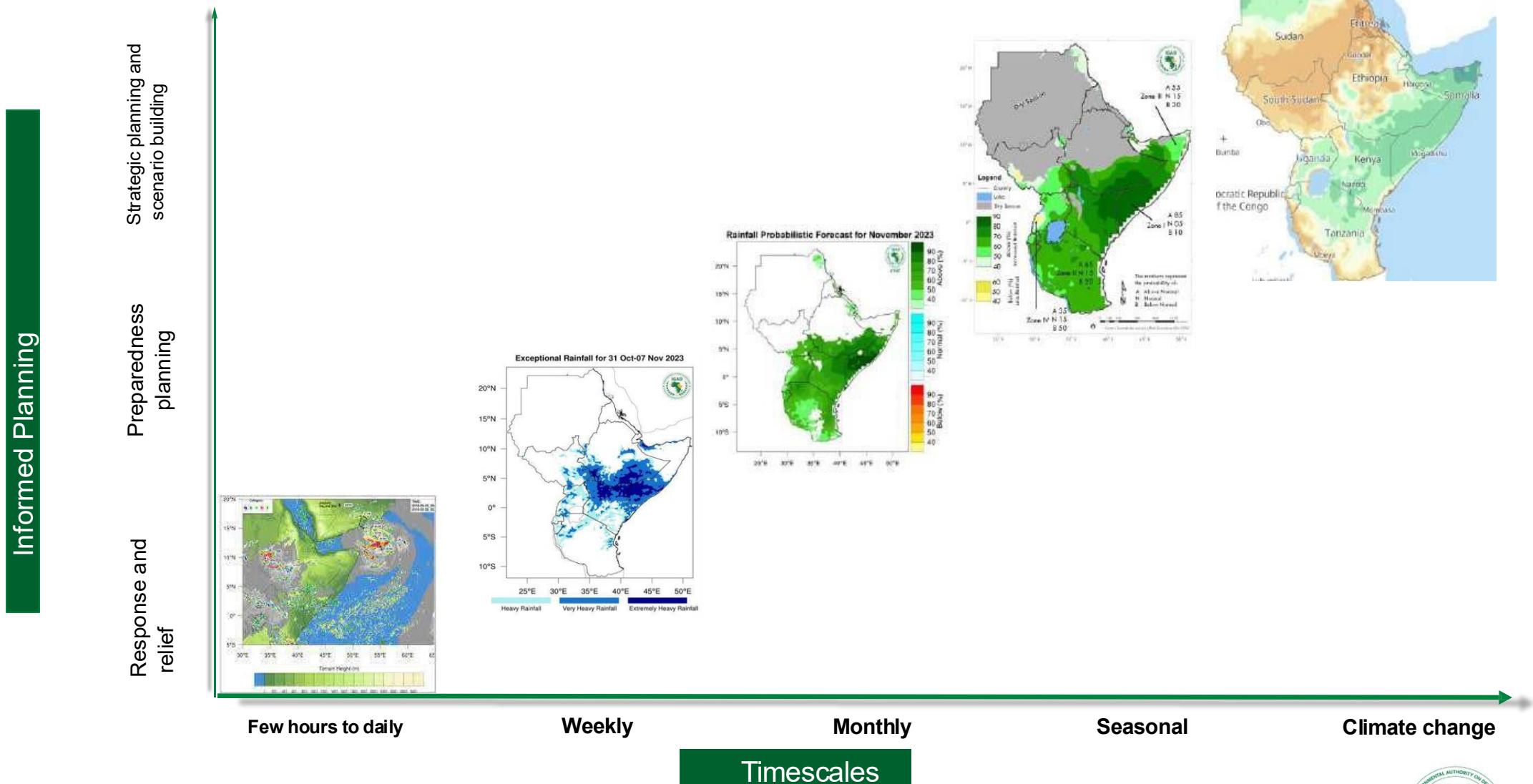
- Produce blended and gridded high resolution regional and national climate data sets using various tools



ICPAC receives data from 150 active weather stations of NMHSs as well as PUMA and RARS (satellite ground receiving stations)

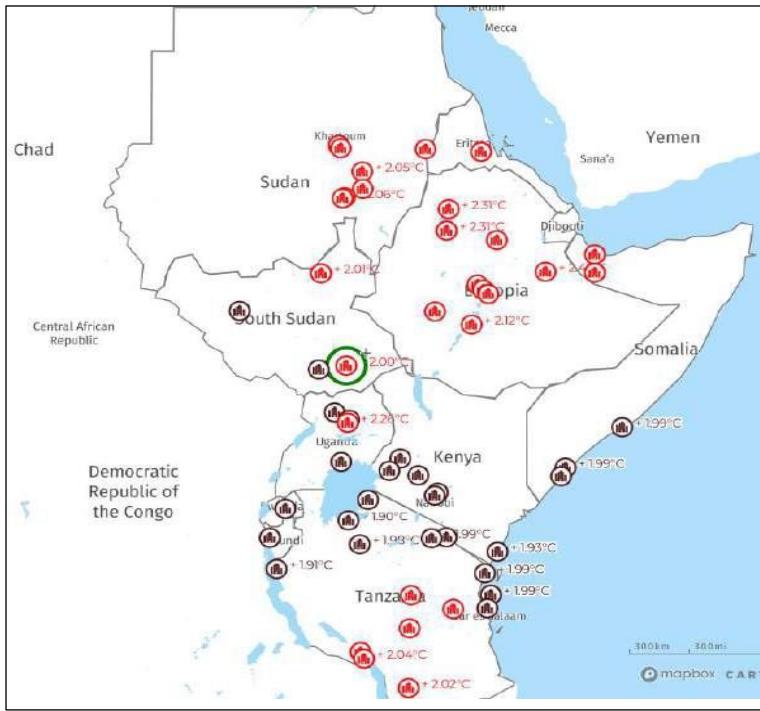


# SEAMLESS EARLY WARNING INFORMATION

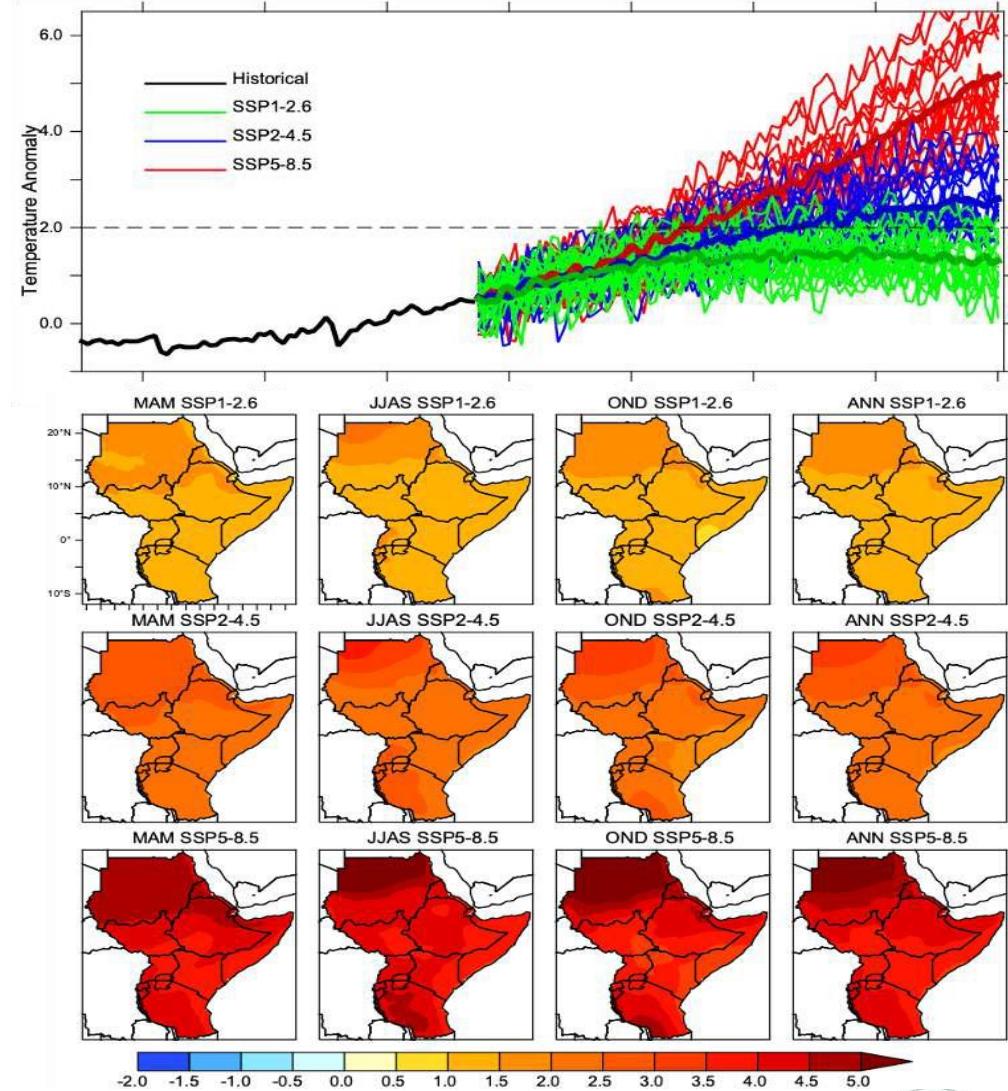


11/17/2025

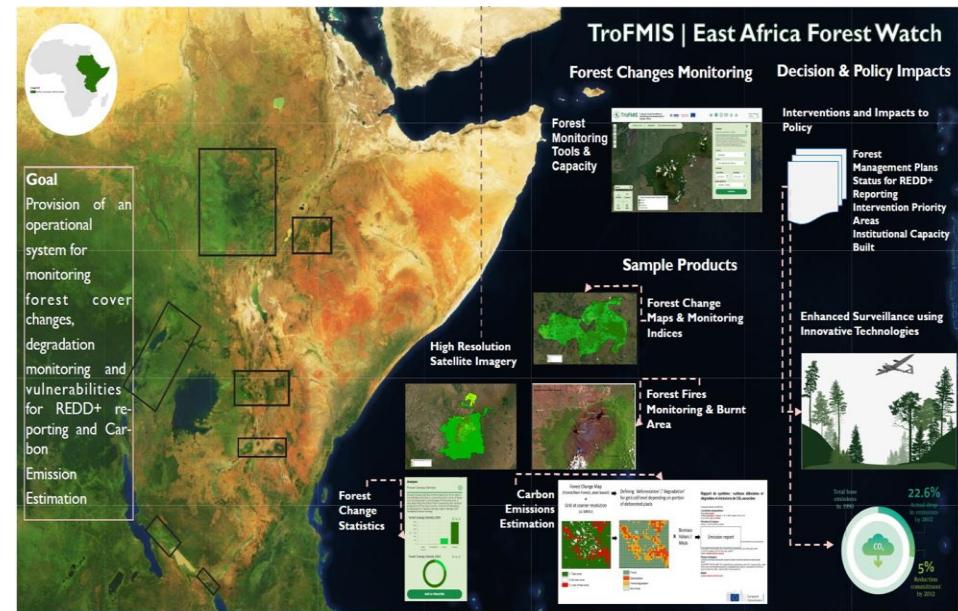
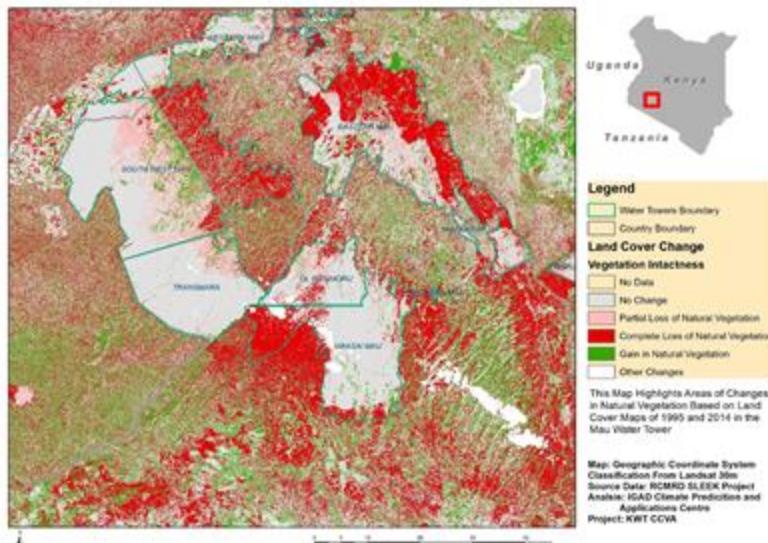
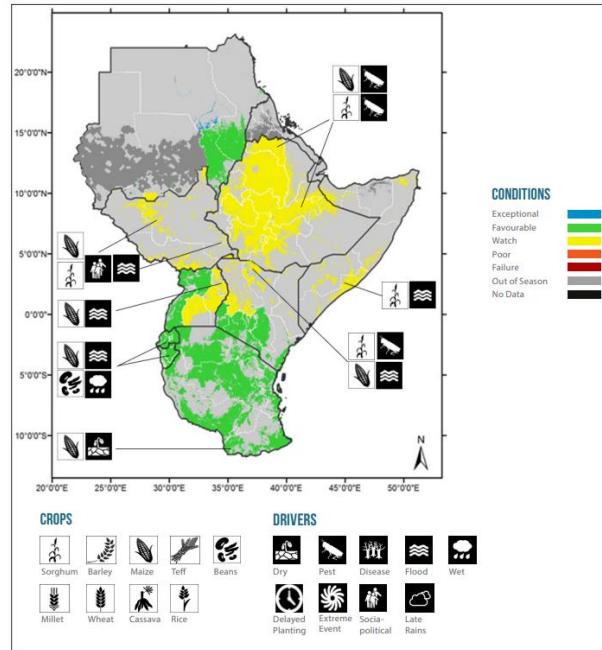
# CLIMATE CHANGE: THE IGAD REGION IS WARMING



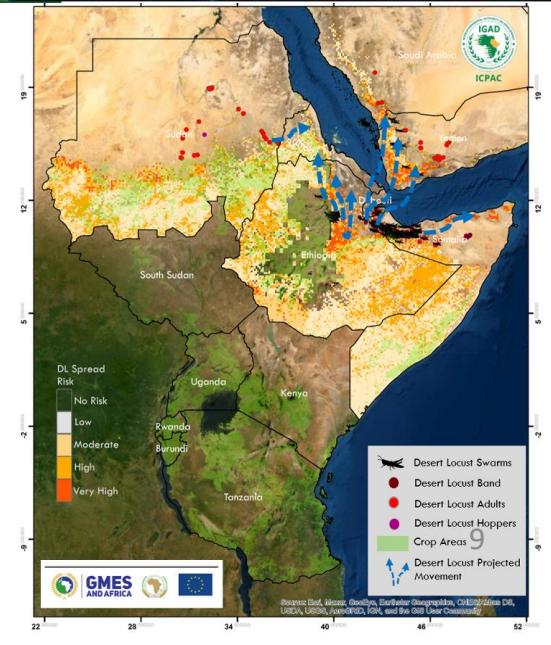
- Eastern Africa has warmed beyond the Paris Agreement agreed value 1.5 and even beyond the 2.0 degree
- Projections shows further warming in the future
- Adaptation and mitigation strategies are key: IGAD climate Change strategy



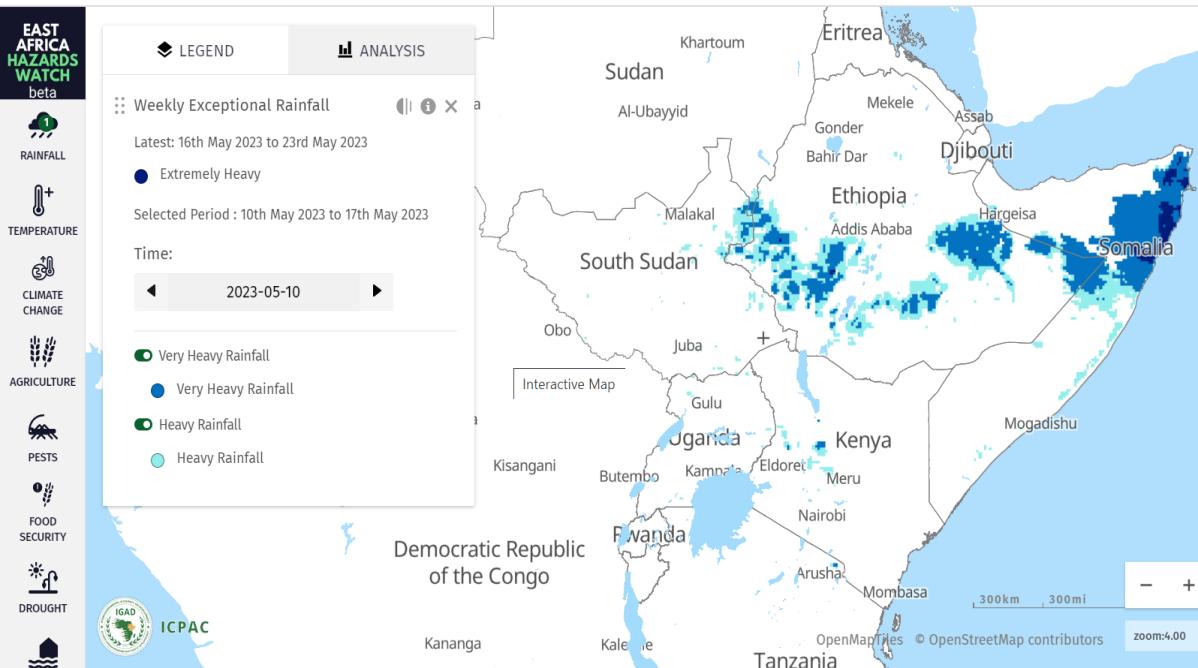
# MONITORING OF CROPS, FORESTS AND PESTS



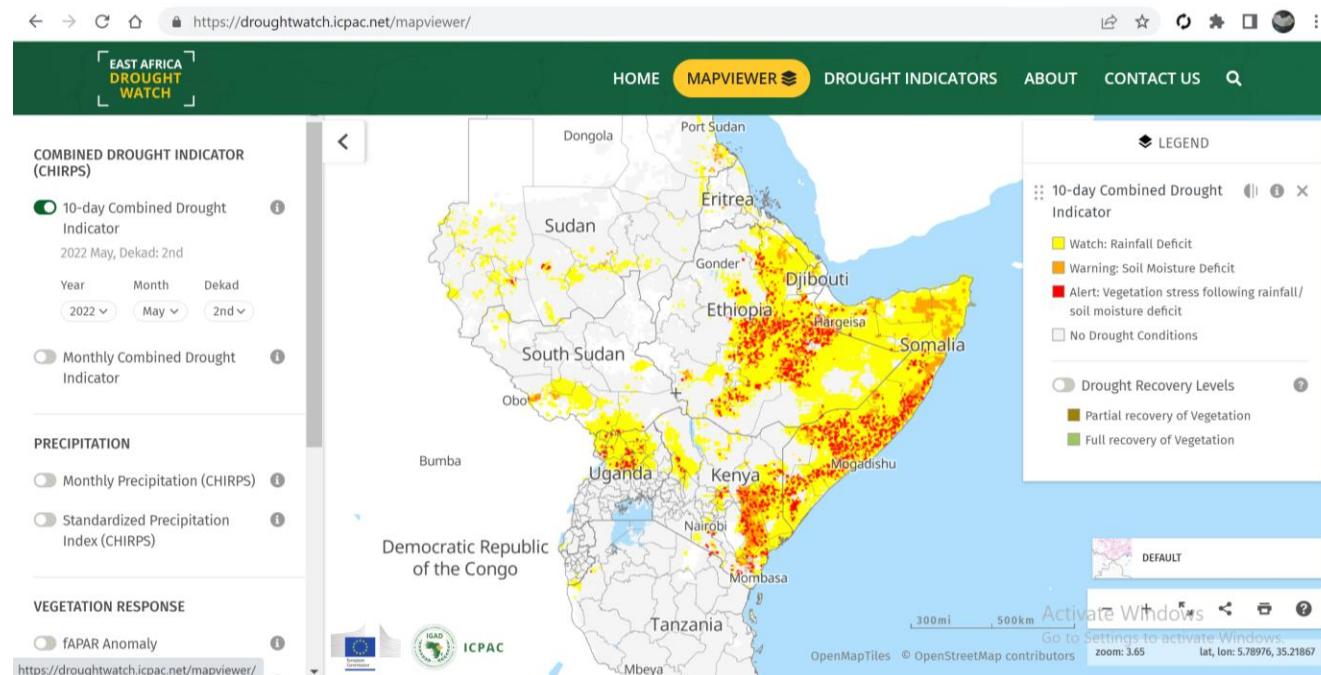
- Crop monitor analyses various crops conditions at different times of the cropping season
- Protected forests monitoring for carbon emission estimation
- Desert locust prediction helps in planning on resources allocation for control measures



# East Africa Hazards watch

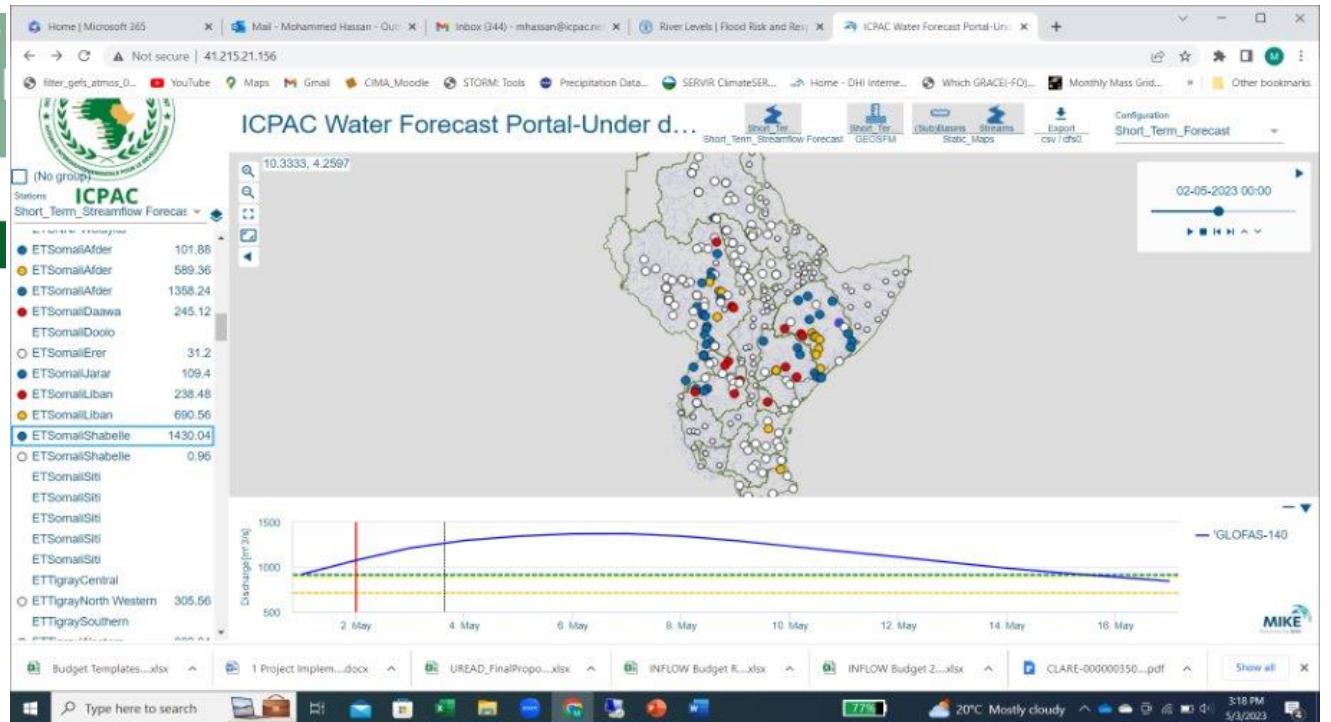
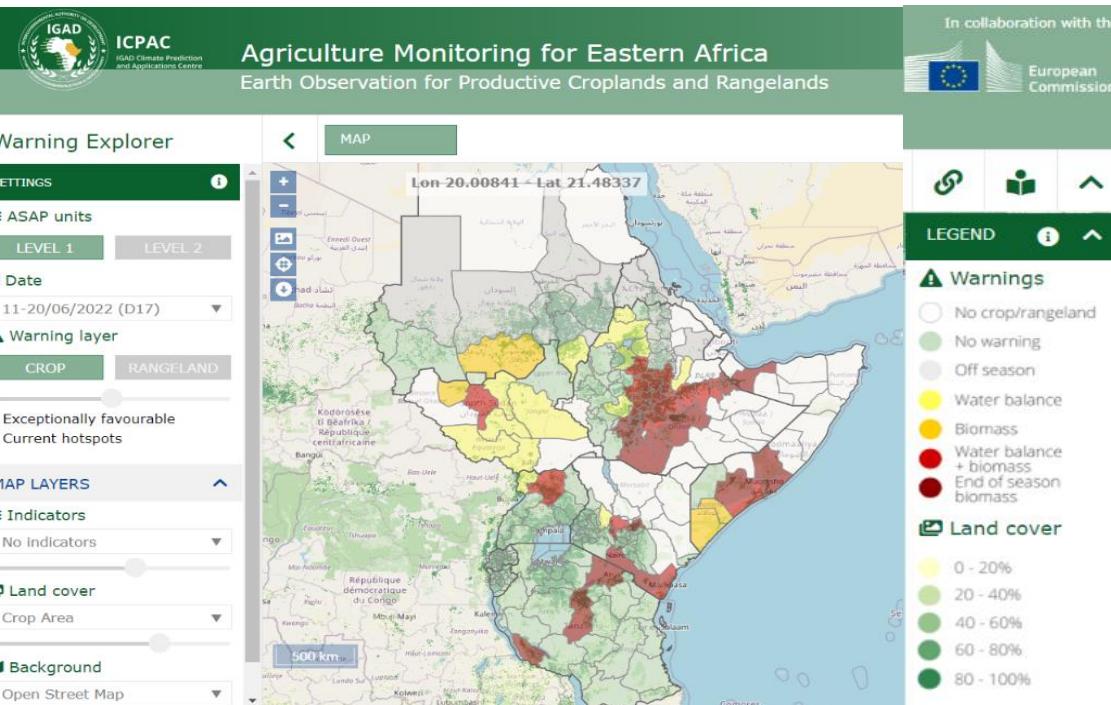


# East Africa Drought watch



# East Africa Agriculture watch

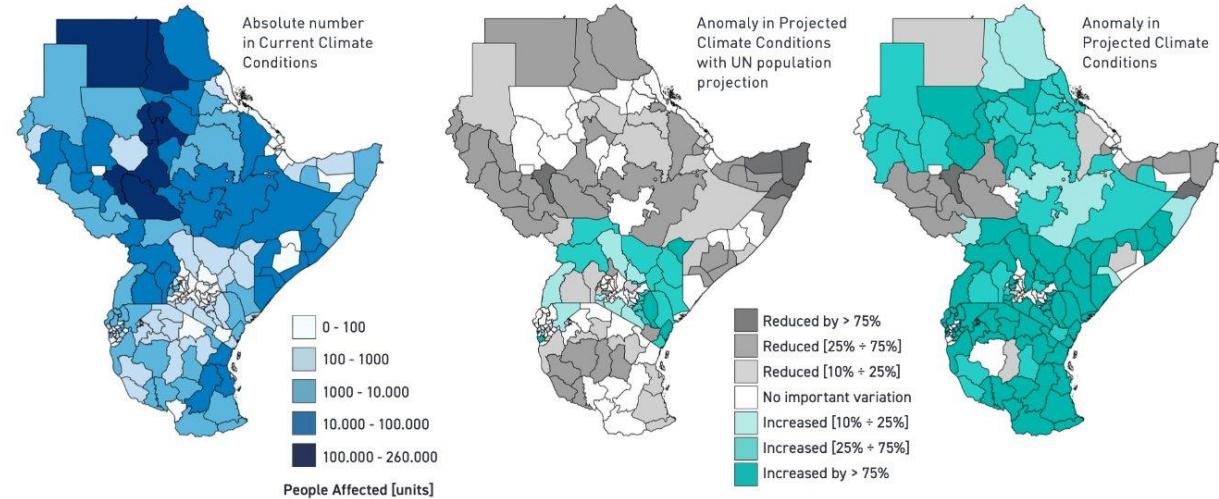
# Water Portal



# REGIONAL RISK ASSESSMENT AND RISK PROFILES

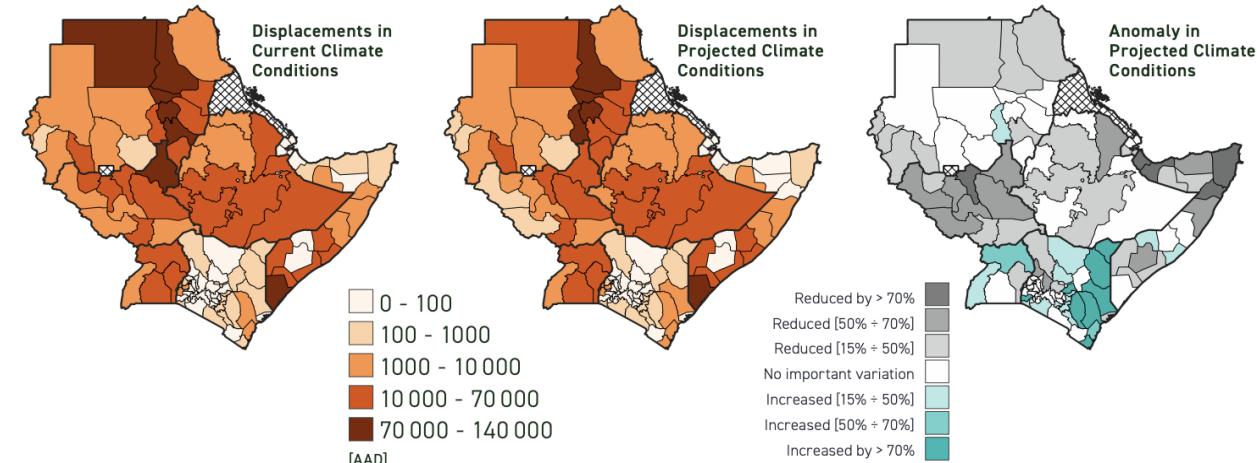
## Flood Risk Profile

*Annual Average Number of People Affected  
by flood events at province or county level - Figure 2*



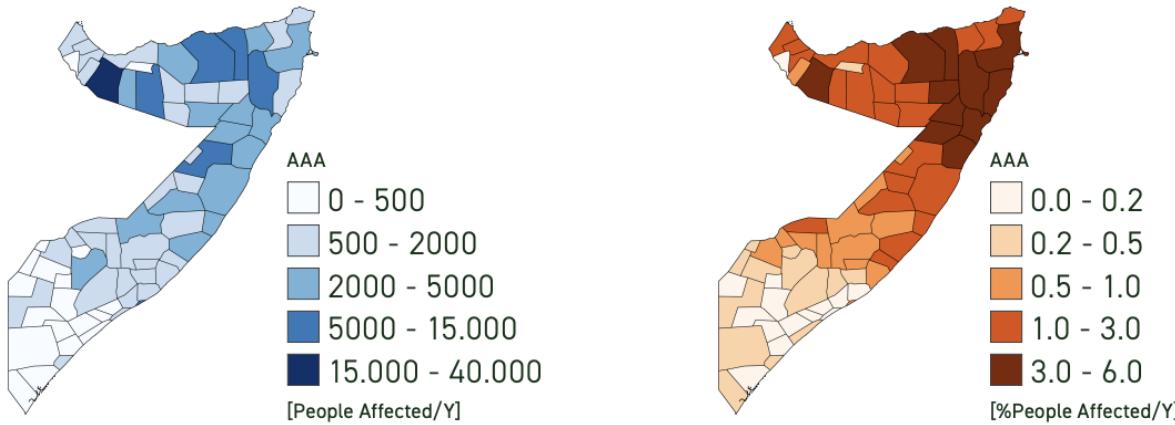
## Displacement /migration modeling

**ANNUAL AVERAGE DISPLACEMENT TRIGGERED BY FLOOD EVENTS AT PROVINCE OR COUNTY LEVEL**



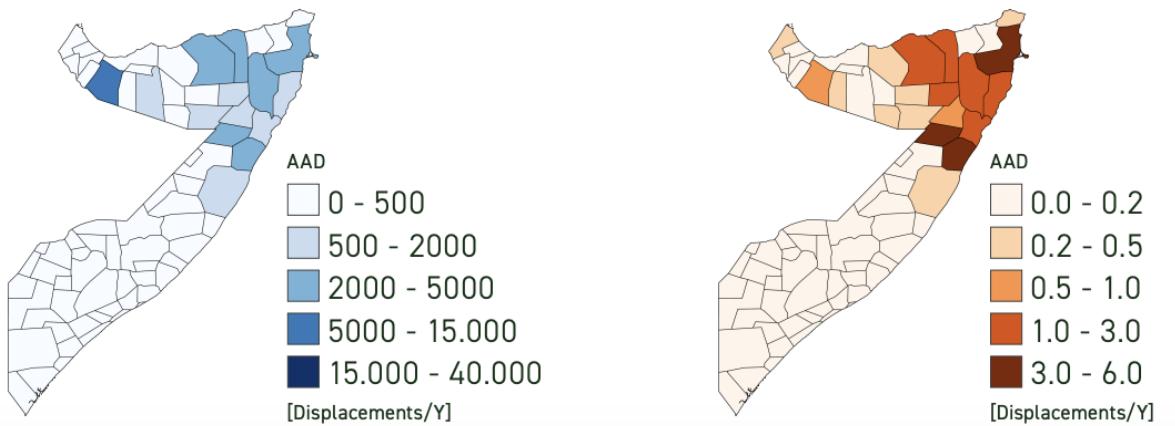
# Cyclone Impacts in Somalia (Loss and Damage)

ANNUAL AVERAGE NUMBER OF PEOPLE AFFECTED BY CYCLONE EVENTS AT DISTRICT LEVEL (ON THE LEFT)  
AND THE CORRESPONDING PERCENTAGE IN RESPECT TO THE TOTAL POPULATION AT DISTRICT LEVEL (ON THE RIGHT)



The total number of people potentially affected by cyclone events in Somalia is around 180,000, corresponding to 1% of the total country population: in some districts in the northeastern part of the country this value exceeds 5%.

ANNUAL AVERAGE NUMBER OF PEOPLE POTENTIALLY DISPLACED BY CYCLONE EVENTS AT DISTRICT LEVEL (ON THE LEFT)  
AND THE CORRESPONDING PERCENTAGE IN RESPECT TO THE TOTAL POPULATION AT DISTRICT LEVEL (ON THE RIGHT)





01

Monitor major hazards and issue early warning information for the region

02

Coordination with national focal institutions on early action.

03

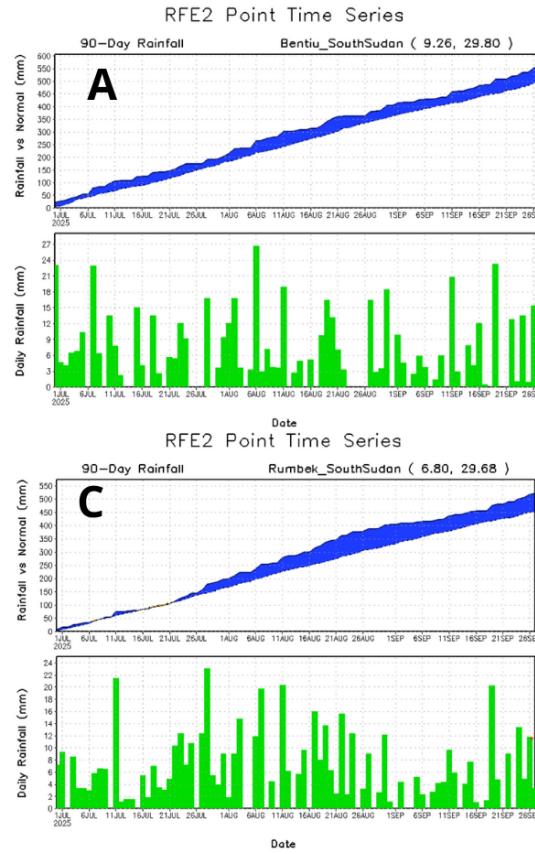
Rapid mapping of affected areas and impacts of disasters

04

Strengthen Capacity to anticipate risk & Support AU Early warning Situation Room

## Disaster Operations Centre - Floods Update

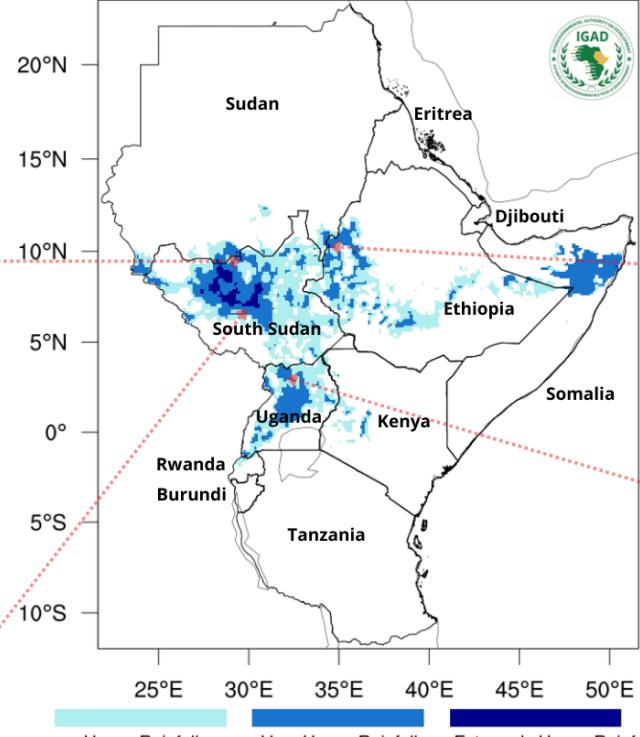
IGAD Multi-Hazard Early Warning Early Action - Situation-Room



### Situation Overview

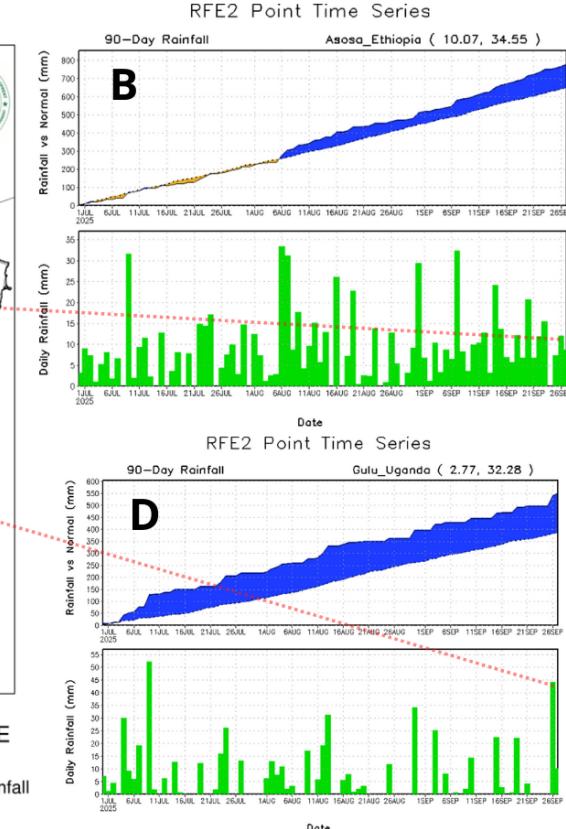
- Some pockets in western Ethiopia, central to north-central South Sudan, and northern Uganda received rainfall in the past week, with daily accumulation in Bentiu, South Sudan, reaching 16mm (A), Asosa, Ethiopia, at 13mm (B), Rumbek, South Sudan at 12mm (C), and Gulu, Uganda, at 44mm (D).

### Exceptional Rainfall for 30 Sep-07 Oct 2025

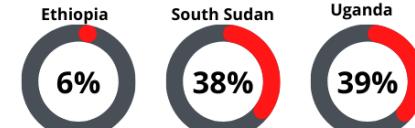


- In the coming week, very heavy rainfall is expected over western Ethiopia, central to north-central South Sudan, and central to north-central Uganda.
- Flooding incidences are expected over the flood prone areas in north-central Uganda, central to north-central parts of South Sudan, and pockets in western Ethiopia.
- Communities in high-risk areas are advised to exercise caution during this period.

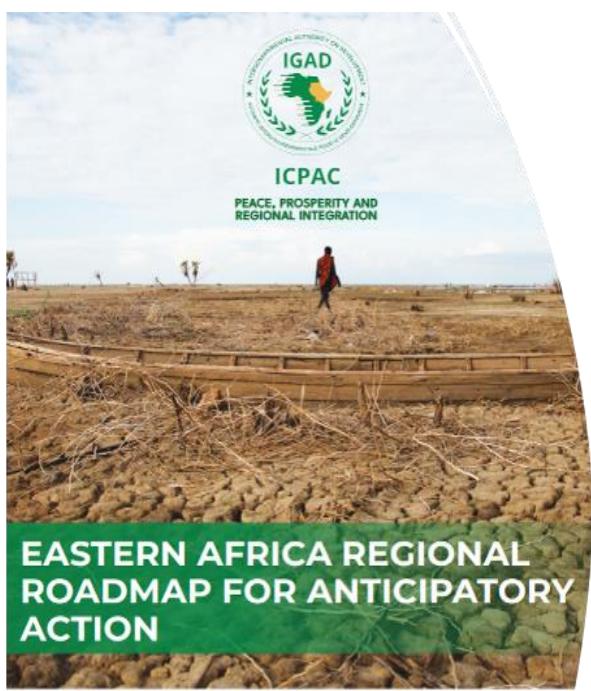
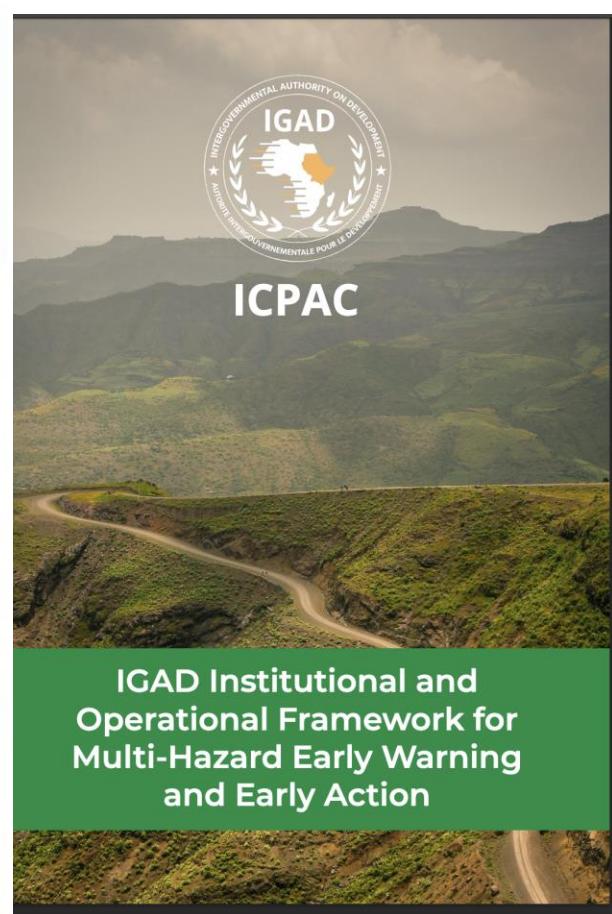
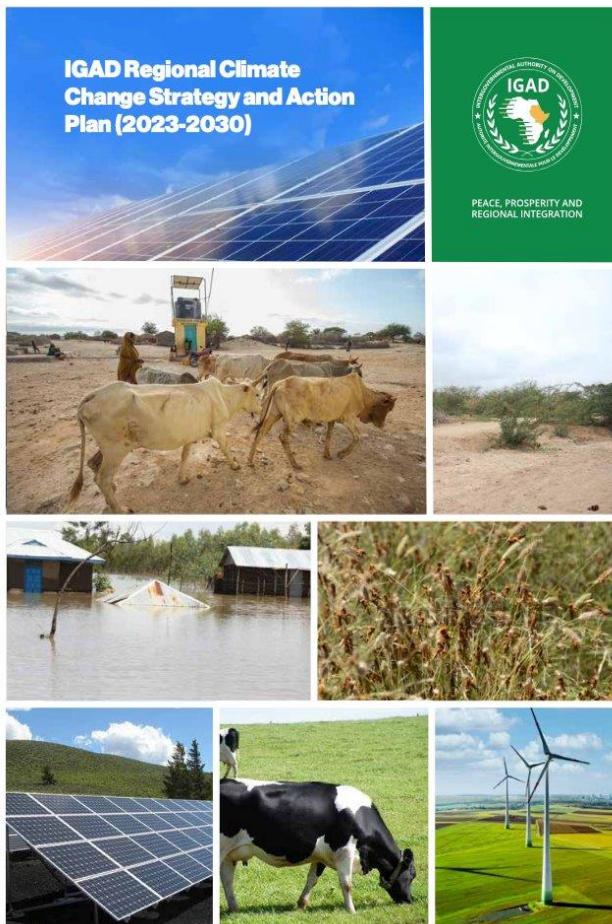
 Issue No.103 September 29, 2025



**Estimated population exposed by countries:**  
In percentage under extremely/very heavy rainfall area

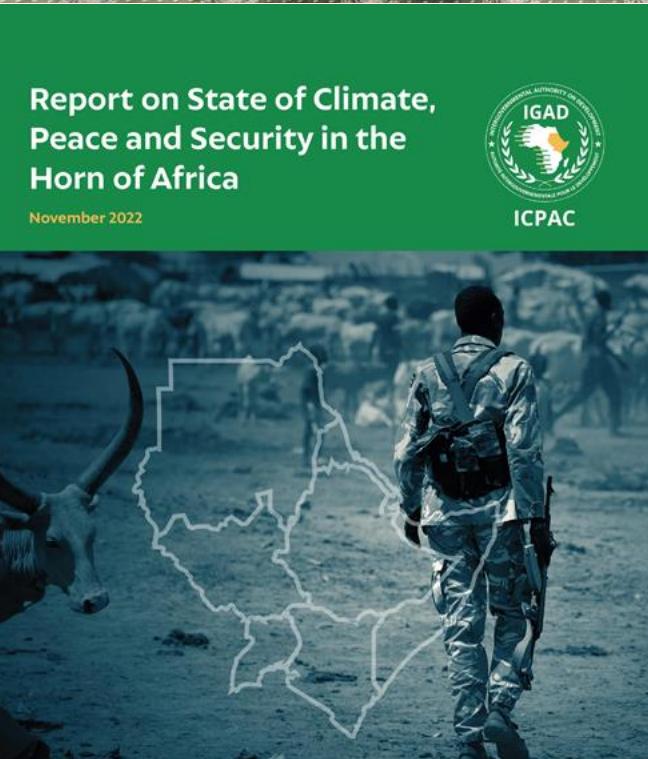
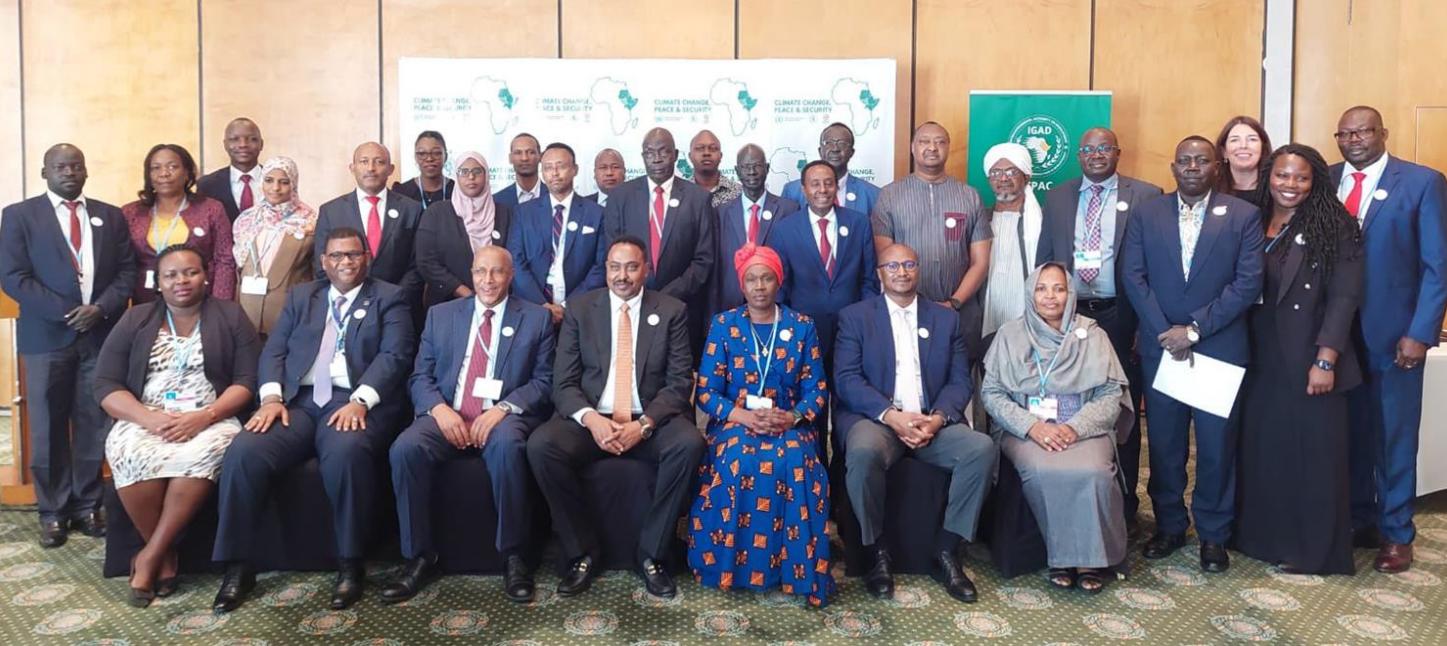


# ICPAC REGIONAL STRATEGIES AND FRAMEWORKS



# IGAD CLIMATE SECURITY COORDINATION MECHANISM

- Climate Security pathways agreed on by technical experts from member states in July 2022;
- Mechanism created by IGAD Sector Ministers at COP 27 in November 2022;
- Endorsed by IGAD Council of Ministers in Nov 2022;
- Endorsed by IGAD Heads of States in June 2023;



# USER ENGAGEMENT: REGIONAL TO COMMUNITY LEVEL



PICSA

## ETHIOPIA

### Disaster Risk Management

The country is expected to experience drought condition affecting the central and northern parts of the country. However, the southern and eastern parts are expected to experience river floods and flash flood with the possibility of wet spells between 4 to 6 days.

#### Advisory

- Destock livestock early.
- Encourage Ministry of Agriculture and Livestock to preserve animal feed.
- Implement drought and flood contingency plan coordinated by Ethiopian Disaster Risk Management Commission (EDRC).
- Clear and improve major drainage systems to be led by the ministry of water.

### Agriculture and Food Security

Expected moisture condition favours crop growth in South-west of Somali region, Southern Oromia and south Ethiopia. Generally good conditions for land preparation and sowing of irrigated farms in the central and northern parts where it is dry climatologically. Wet conditions in the south favours breeding of Desert locust and other pests. Likelihood of flash flooding in some part of southern Ethiopia which might impact on croplands.

#### Advisory

- Provide agricultural inputs on time to OND benefiting areas of southern and eastern Ethiopia.
- Issue early warning alerts and heightened measure against possible desert locust invasion.
- Undertake proper post-harvest management for the main season crop (JJA).

### Water and Energy

Adequate water availability in areas where the OND is the rainy season. Risk of flooding in downstream areas.

#### Advisory

- Raise awareness on flood risk areas in time.
- Continuously monitor dam levels.

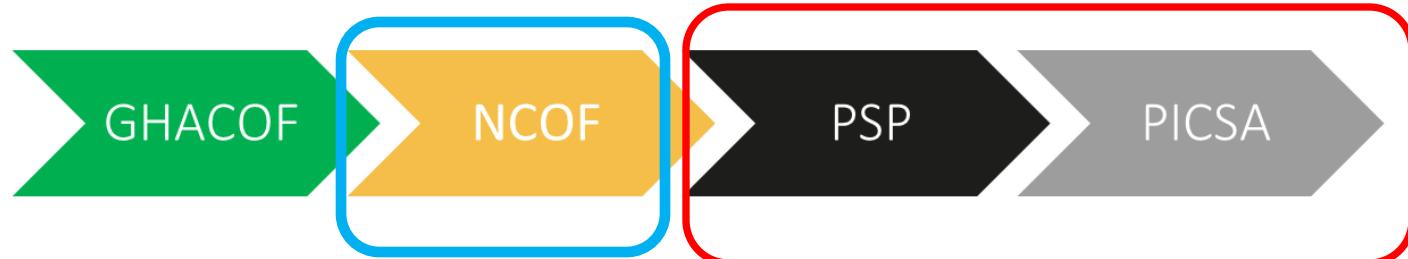
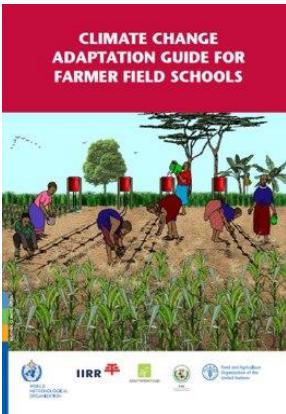
### Livestock

Regeneration of pasture, availability of water, reduce livestock mobility in search of water and pasture. Food security and good nutrition. Reduced pneumonia due to predicted warmer temperatures. Increased livestock productivity (meat, milk, hides & skins). Expected stable prices or improved due to exports and good animal body condition. Favorable for vaccination and water harvesting.

#### Advisory

- Enhance disease surveillance especially for RVF.
- Enhance production and conservation of fodder including benefits from JJAS beneficiary areas.
- Encourage vaccination of animals.
- Facilitate community awareness about expected enhanced rains either to relocate due to floods, plant fodder, present animals for vaccination, water harvesting etc.
- Increase control efforts towards desert locust.

# Linking Regional to Community Level



NCOFs are held at national level and a summary for decision makers produced for policy makers

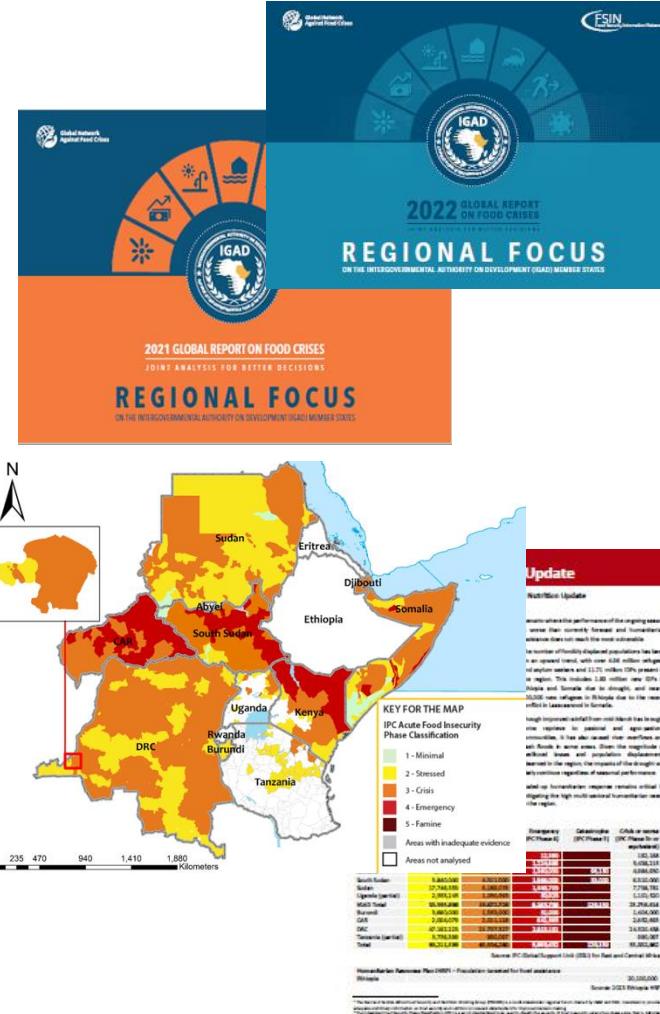
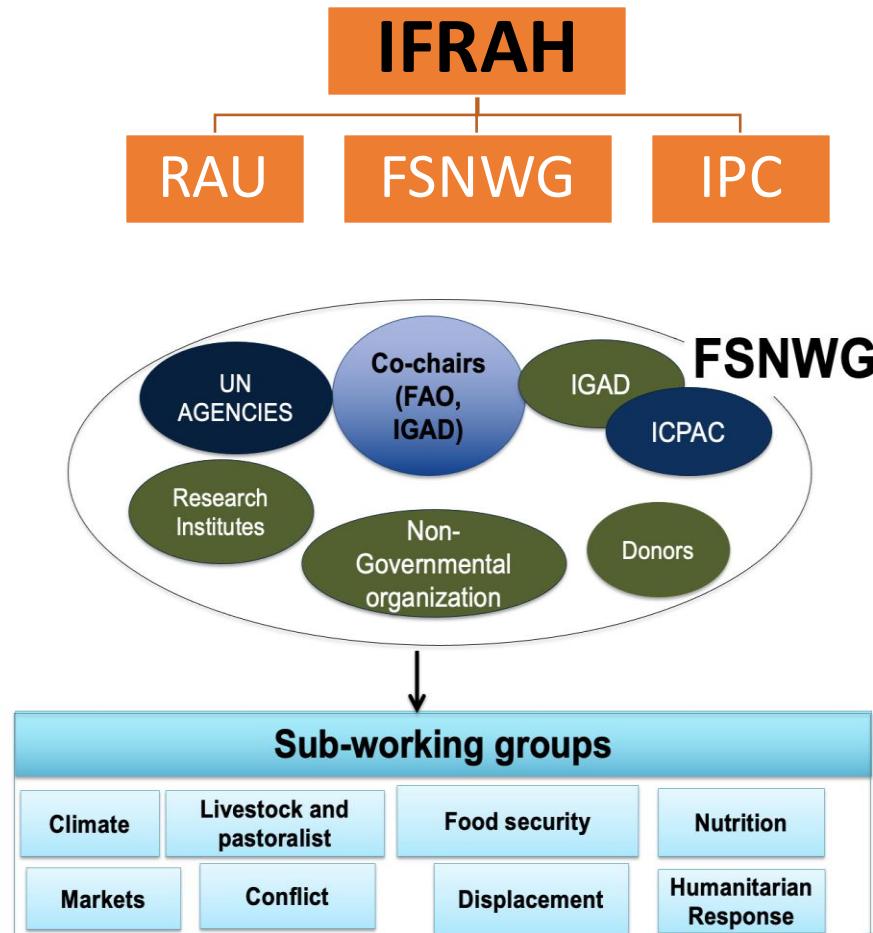
Stakeholders generating agricultural advisories based on seasonal forecast during a PSP event

Small holder  
farmer women  
trained on  
PICSA

# CLIMATE CHANGE ADAPTATION INITIATIVES (GARISSA-KENYA)



# IGAD Food security, nutrition & resilience analysis hub





## CAPACITY BUILDING TRAININGS FOR MEMBER STATES

**Key Areas of Capacity Building:** climate modelling and prediction, co-production, climate change adaptation, risk assessment, Early warning systems, anticipatory action, DRR, communication, ..etc.



# Some selected high impacts Publications

[View PDF](#) [Download full issue](#)

Original research article

## Increasing the prospective capacity of global crop and rangeland monitoring with phenology tailored seasonal precipitation forecasts

Michele Meroni <sup>a,1</sup>, Petar Vojnović <sup>b,1</sup>, Matteo Zampieri <sup>c, d</sup>, Stefano Materia <sup>e, f</sup>, Felix Rembold <sup>c</sup>, Oliver Kipkogei <sup>g</sup>, Andrea Toreti <sup>c</sup>

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<https://doi.org/10.1016/j.ciser.2023.100434> [Get rights and content](#)

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### Abstract

Droughts are more and more often a limiting factor to agricultural production and can have severe negative effects on food security in

## Projected future changes in food insecurity hotspots over the IGAD region of Eastern Africa

Original Paper | [Open access](#) | Published: 14 August 2024

Volume 17, article number 245, (2024) [Cite this article](#)

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Paulino Omoj Omay , Josiah M. Kinama, Nzioka J. Muthama, Christopher Oludhe, Guleid Artan & Zachary Atheru

 563 Accesses [Explore all metrics](#) 

[Open Access](#) [Technical Note](#)

## New Functionalities and Regional/National Use Cases of the Anomaly Hotspots of Agricultural Production (ASAP) Platform

by Felix Rembold <sup>1,\*</sup>, Michele Meroni <sup>1</sup>, Viola Otieno <sup>2</sup>, Oliver Kipkogei <sup>2</sup>, Kenneth Mwangi <sup>2</sup>, João Maria de Sousa Afonso <sup>3</sup> , Isidro Metódio Tulem Johannes Ihadua <sup>3</sup> , Amílcar Ernesto A. José <sup>3</sup> , Louis Euvène Zoungnana <sup>4</sup> , Amjed Hadj Taieb <sup>4</sup> , Ferdinando Urbano <sup>1</sup> , Maria Dimou <sup>1</sup> , Hervé Kerdiles <sup>1</sup> , Petar Vojnović <sup>1</sup> , Matteo Zampieri <sup>1,5</sup>  and Andrea Toreti <sup>1</sup> 

<sup>1</sup> Joint Research Centre (JRC), European Commission, Via E. Fermi 2749, I-21027 Ispra, VA, Italy

<sup>2</sup> Intergovernmental Authority on Development (IGAD) Climate Prediction and Applications Centre (ICPAC), Nairobi P.O. Box 10304-00100, Kenya

<sup>3</sup> Instituto Nacional de Meteorologia e Geofísica de Angola (INAMET), Rua 21 de Janeiro, Rotunda do Gamek à Direita, S/N-RC, Luanda 1323, Angola

<sup>4</sup> GIS and Remote Sensing Unit, Observatoire du Sahara et du Sahel (OSS), P.O. Box 31, Boulevard du Leader Yasser Arafat, Tunis 1080, Tunisia

<sup>5</sup> Department of Physical Science and Engineering, King Abdullah University of Science and Technology (KAUST), Thuwal 23955, Saudi Arabia

\* Author to whom correspondence should be addressed.

*Remote Sens.* **2023**, *15*(17), 4284; <https://doi.org/10.3390/rs15174284>

Submission received: 26 June 2023 / Revised: 21 July 2023 / Accepted: 5 August 2023 /

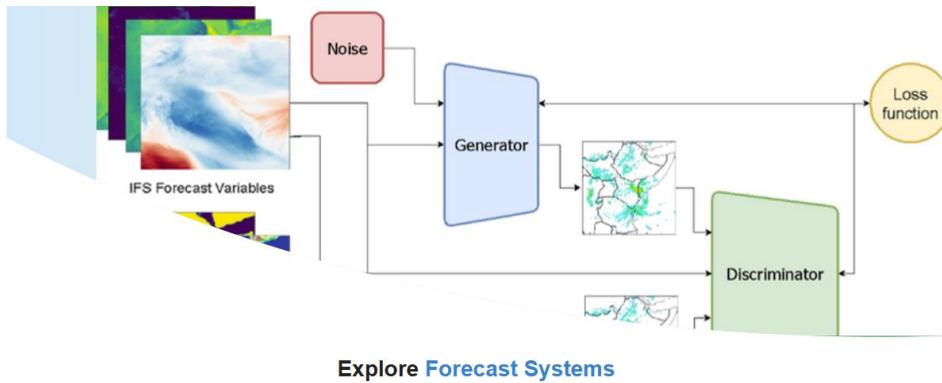
Published: 31 August 2023

TITLE	AUTHORS	JOURNAL
Increasing the prospective capacity of global crop and rangeland monitoring with phenology tailored seasonal precipitation forecasts.	Meroni, M., et al.	Climate Services <a href="https://doi.org/10.1016/j.ciser.2023.100434">https://doi.org/10.1016/j.ciser.2023.100434</a>
Atmospheric and oceanic conditions associated with early and late onset for Eastern Africa short rains.	Gudoshava, M. et al.	International Journal of Climatology <a href="https://doi.org/10.1002/joc.7627">https://doi.org/10.1002/joc.7627</a>
Forage Monitoring and Prediction Model for Early Warning Application over the East of Africa Region	Ouma, J et al.	Journal of Atmospheric Science Research <a href="https://doi.org/10.30564/jasr.v5i4.4809">https://doi.org/10.30564/jasr.v5i4.4809</a>



# Innovation

- **Machine Learning (ML) and Artificial Intelligence (AI) to Strengthening Early Warning Systems for Anticipatory Action (SEWAA)**



This section provides quick visualization and exploration options for various weather and climate forecasting models.



#### cGAN 1000 Ensemble

State-of-the-art AI-based weather models to give more accurate and local predictions enabling particular countries and regions to better anticipate and prepare for extreme weather.



#### cGAN 50 Ensemble

State-of-the-art AI-based weather models to give more accurate and local predictions enabling particular countries and regions to better anticipate and prepare for extreme weather.



#### ECMWF IFS Ensemble Forecast

The IFS is a sophisticated data assimilation system combined with a global numerical model of the Earth system, supported by infrastructure to make forecast products available to Member and Co-operating States, and other users.



#### Global Ensemble Forecast System (GEFS)

A weather forecasting model developed by the National Centers for Environmental Prediction (NCEP). It generates 21 separate forecasts, or



#### Embedded External System

Forecast products sourced from partner systems and embedded as-is.

- **Machine Learning (ML) and Artificial Intelligence (AI) to enhance Risks knowledge and Impact Based Forecasting**

**E4DRR**

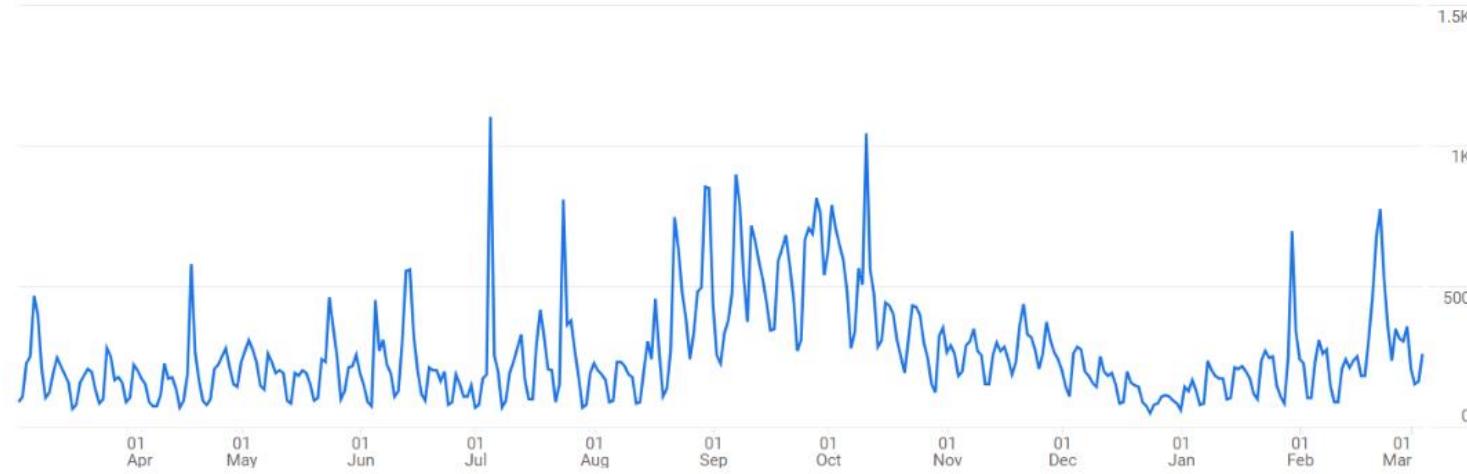
Risk knowledge is an essential component of effective disaster risk management. The available impact data from previous disasters serve as invaluable tools for disaster preparedness and decisionmaking support. The project, titled "Hazard Modeling, Impact Estimation, Climate Storylines for Drought and Flood Disasters in Eastern Africa" (in short form E4DRR), is a two-year project funded by the Complex Risk Analytics Fund (CRAF) and implemented by ICPAC in collaboration with NORCAP.

### Aim and Objectives

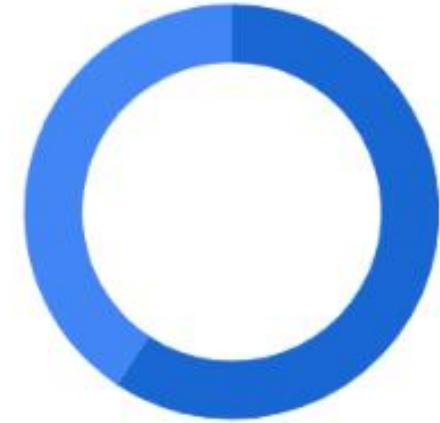
To enhance the East Africa Hazard Watch Portal as a decision-making and actionable information tool for Disaster Risk Management (DRM) through impact-based forecasting based on a chain of auditable evidence synthesised from event-based climate Storylines.

- Enhanced Risk knowledge**  
Enhanced Risk Knowledge and Awareness Through the Creation of Event-Based Climate Storylines
- Operationalizing Impact Based Forecasting**  
Exploration of storylines creation processes and applications in Ensemble Prediction System (EPS) Impact Based Forecasting (IBF)
- Capacity Development**  
On the use of methods and tools for storyline creation

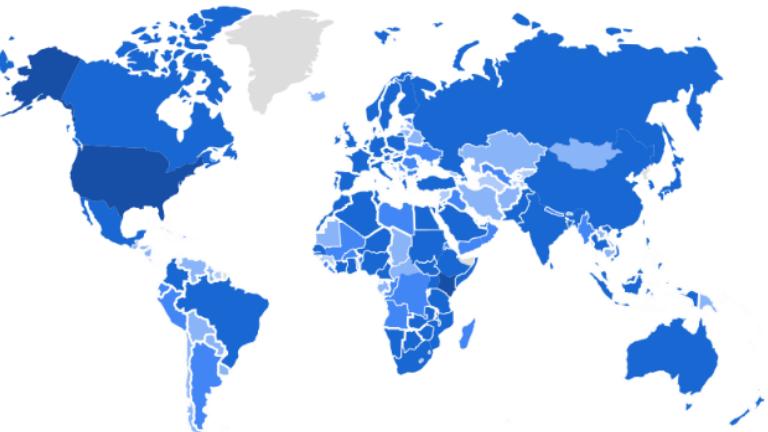
### Users by Audience name over time



### Users by Gender



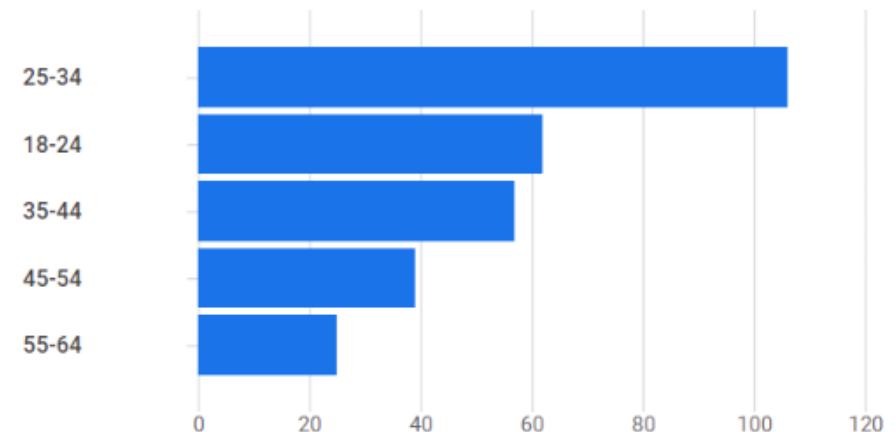
### Users by Country



COUNTRY	USERS
Kenya	26K
United States	7K
Somalia	4.8K
United Kingdom	3.5K
Ethiopia	3.3K
Netherlands	3.3K
Uganda	2.9K

[View countries →](#)

### Users by Age





~ 300 million  
people

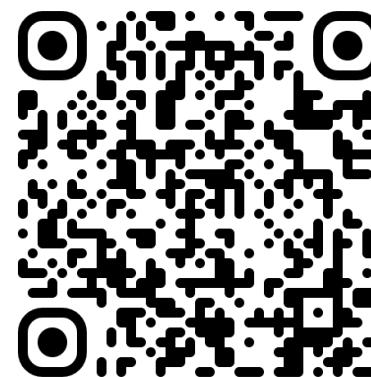


## Multi-Channel On-Boarding

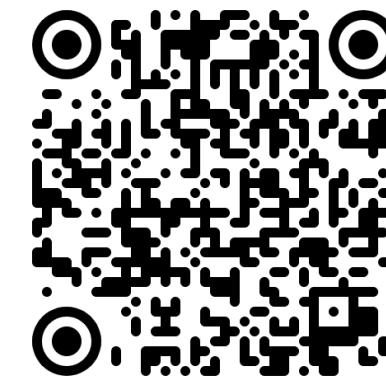


Web Platform

**HUSIKA.ICPAC.NET**



Android OS



iPhone/iPad



**ICPAC**

**Thank you !**



# HUMAN MOBILITY IN THE CONTEXT OF DISASTERS & CLIMATE CHANGE

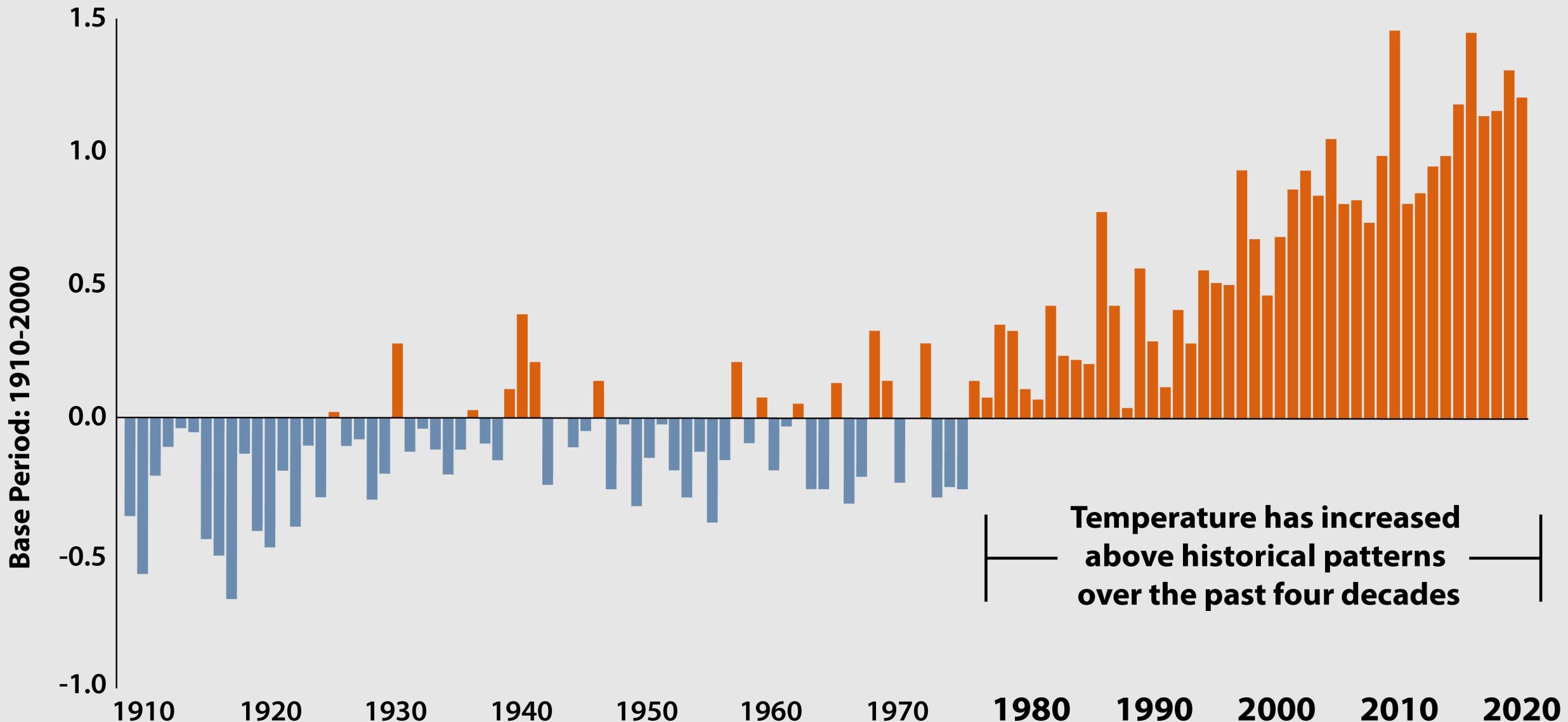
IGAD Climate Prediction and Applications Centre (ICPAC)



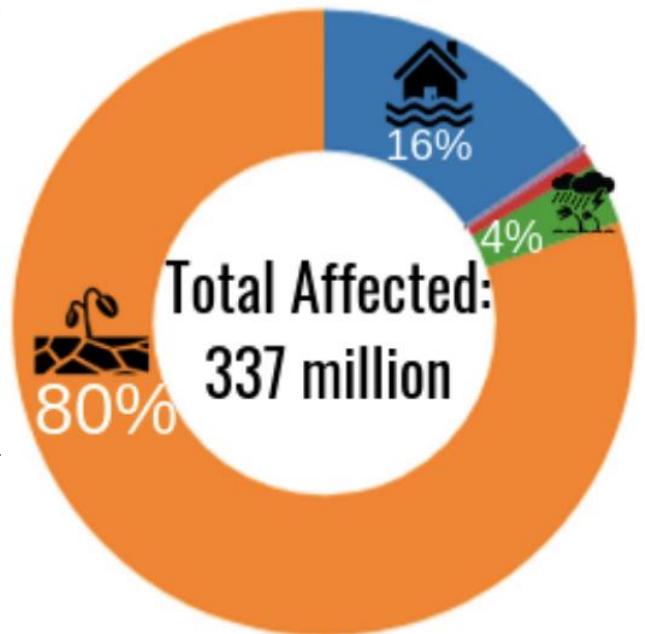
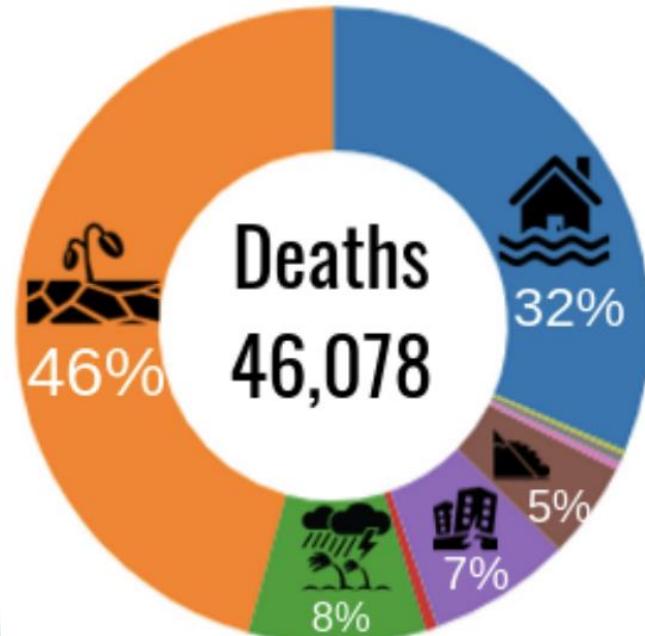
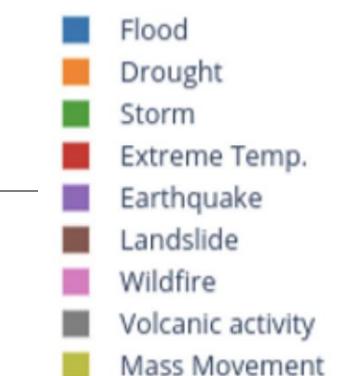
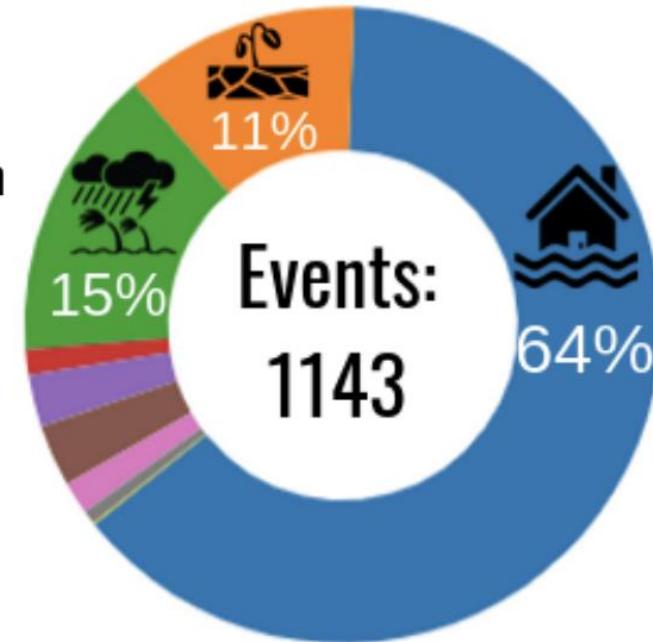
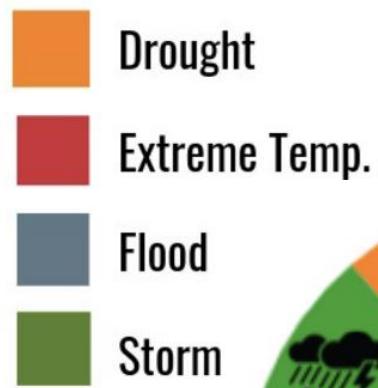
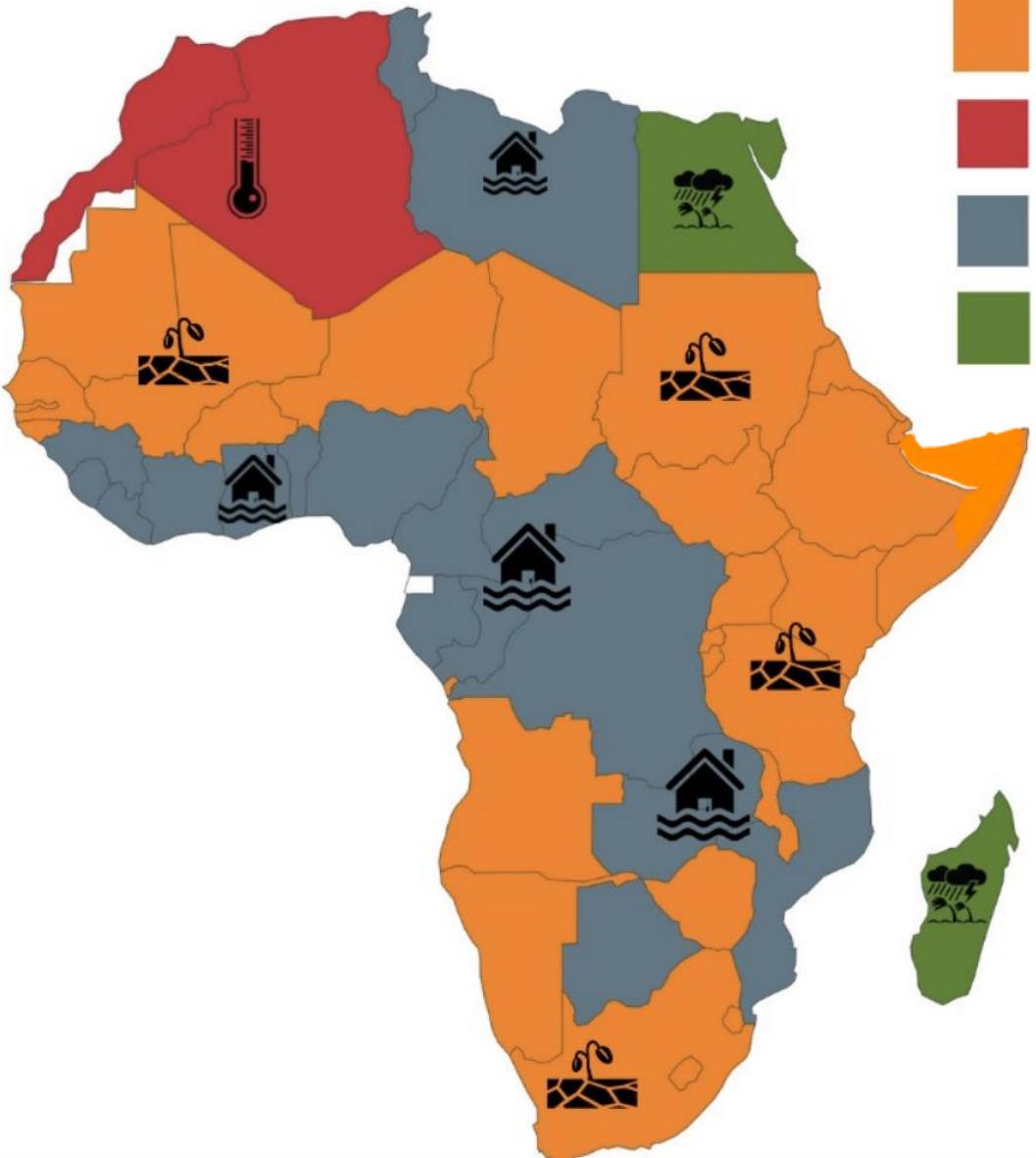
Implemented by:



# Africa's Temperature Anomalies, 1920-2020 (degrees Celsius)



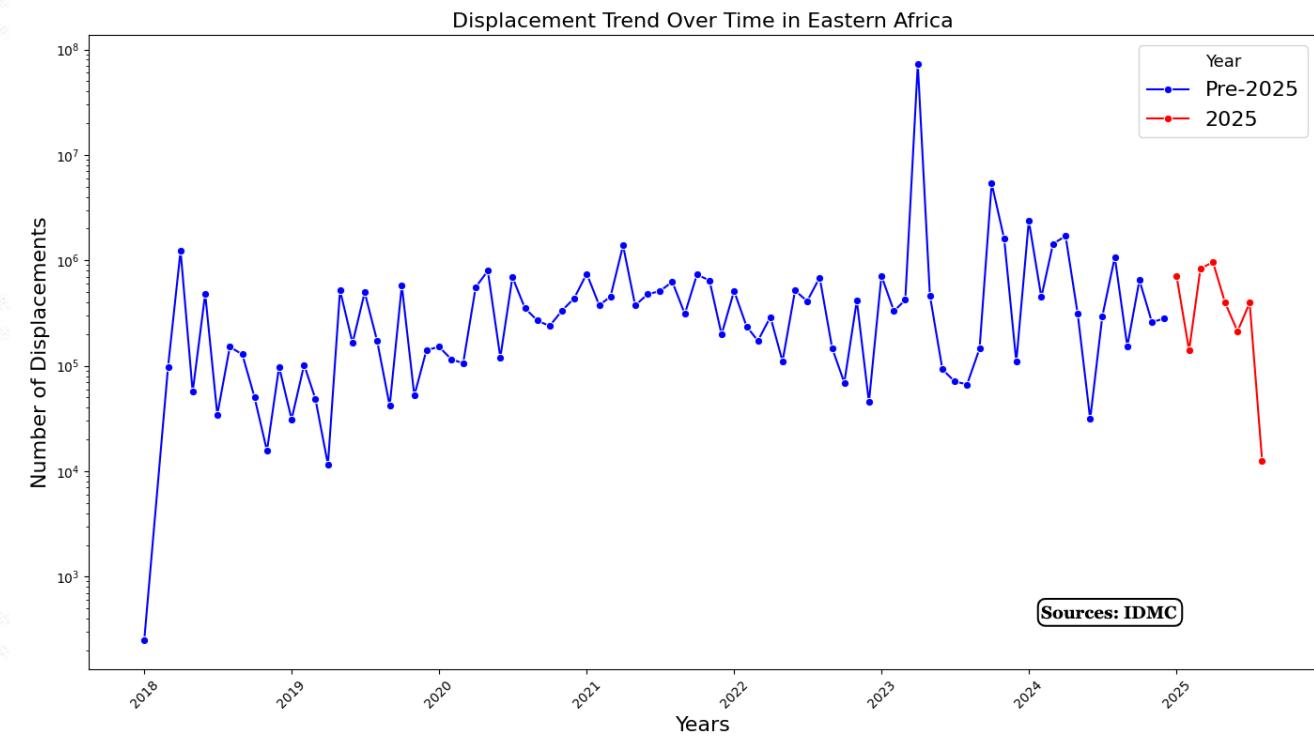
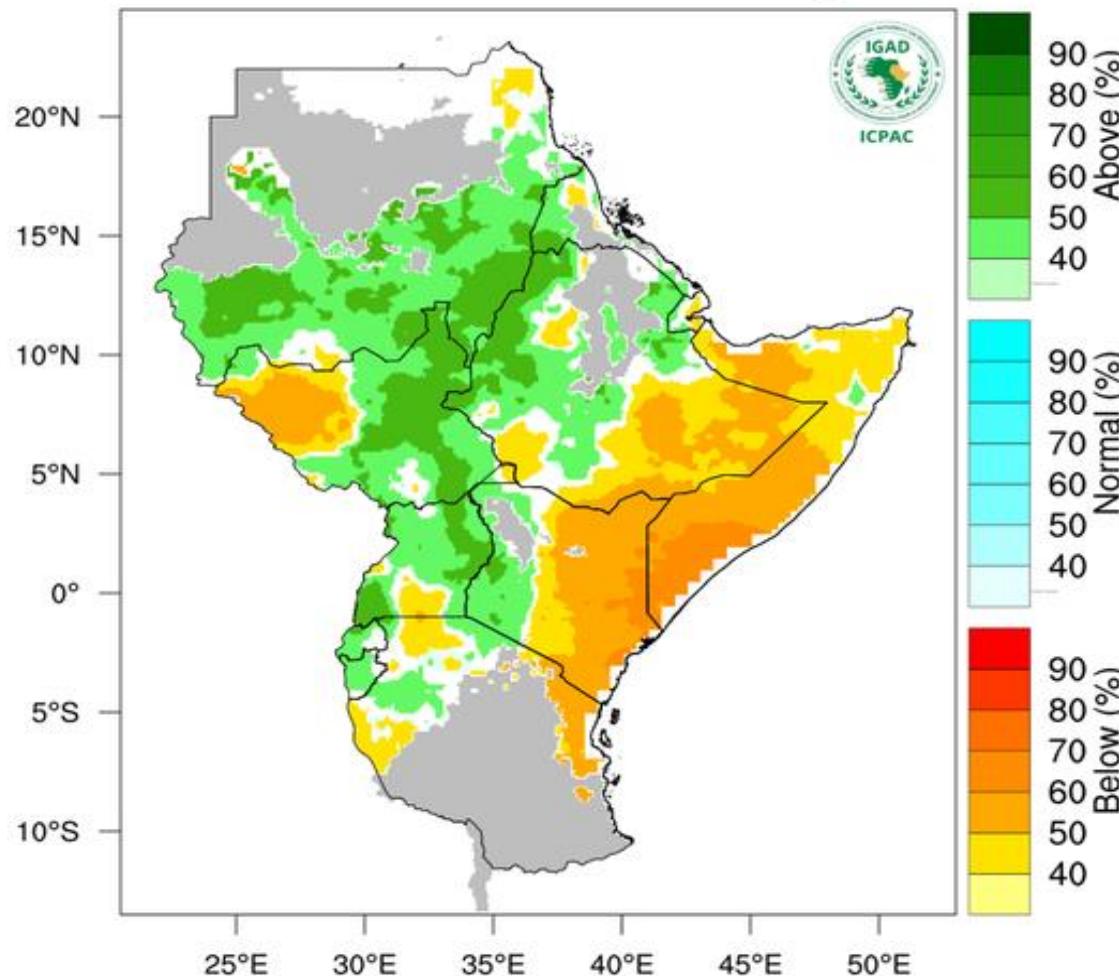
## Disaster Type Affecting Highest Number of People by Country (2000-2019\*)



# CONTEXT AND RATIONALE

- IGAD is among Africa's most climate-vulnerable regions.
- 3.5 million people displaced in 2023 by droughts & floods.
- Climate, disasters, and conflicts coupled with weak governance constrain development, especially in borderlands.
- Frameworks exist (IGAD DRM, climate change, IGAD Migration policy, Free Movement Protocol, Kampala MECC, Transhumance Protocol...), but domestication & implementation remain limited.
- Disaster displacement remains a pressing reality that continues to disrupt lives and livelihoods, demanding urgent and coordinated action.

## Rainfall Probabilistic Forecast for Sep – Nov 2025



# MODIAC PROJECT : GOAL AND OBJECTIVES

Goal: to build resilience and support safe, dignified mobility across the IGAD region.

Objectives: To strengthen IGAD and Member States' capacity to anticipate, plan, and manage mobility linked to climate change and disasters

Core areas:

- Climate–mobility data modelling and forecasting ( ICPAC)
- Integration into early warning and planning systems
- Policy harmonisation and cross-border dialogue fora
- Gender-responsive and community-based approaches



# MODIAC OUTPUTS

Output1-Strengthening IGAD's  
Technical Role

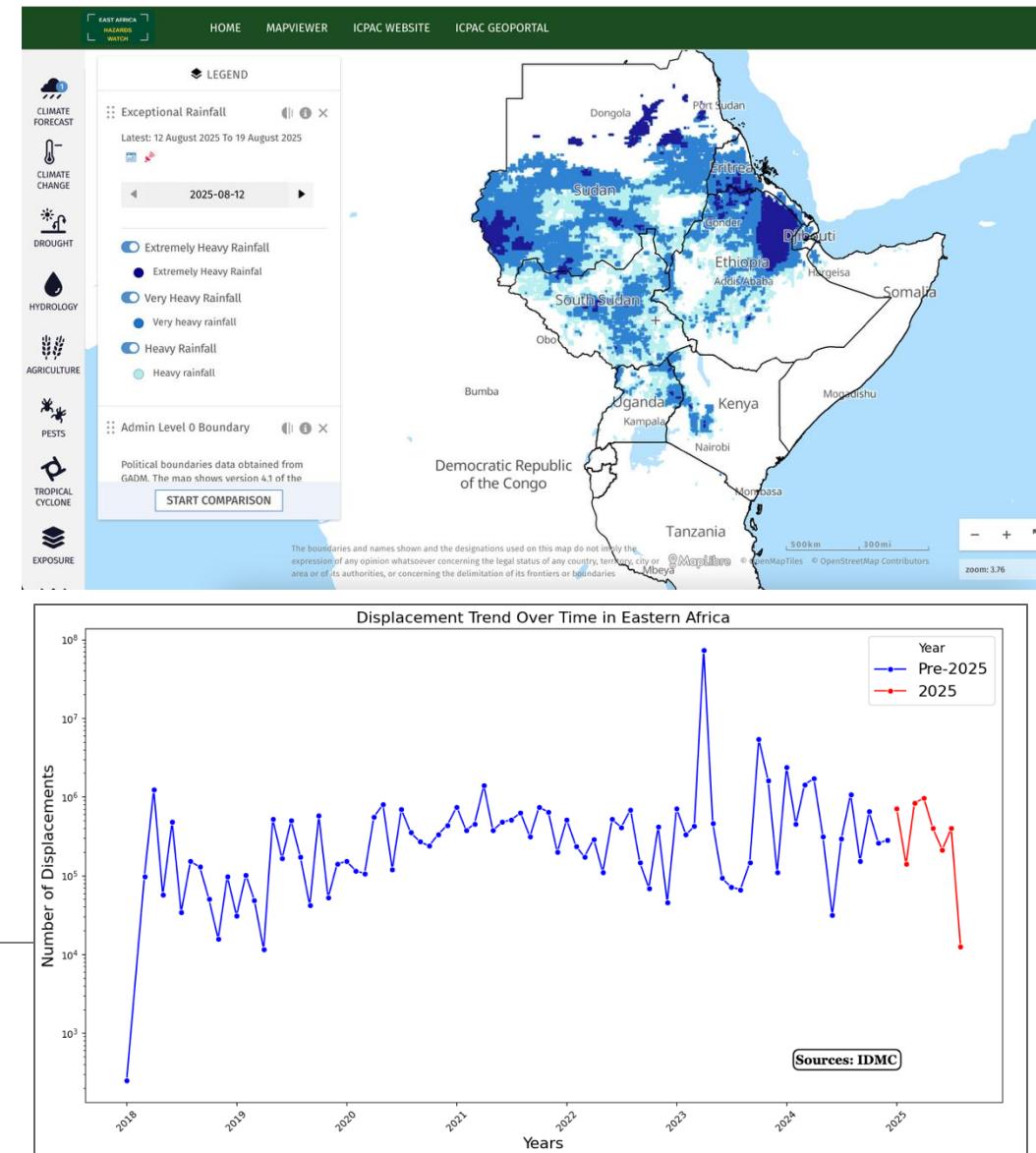
Output 2-Enhancing ICPAC's  
Analytical Capacity

Output3- Empowering Local Action  
in Border Areas

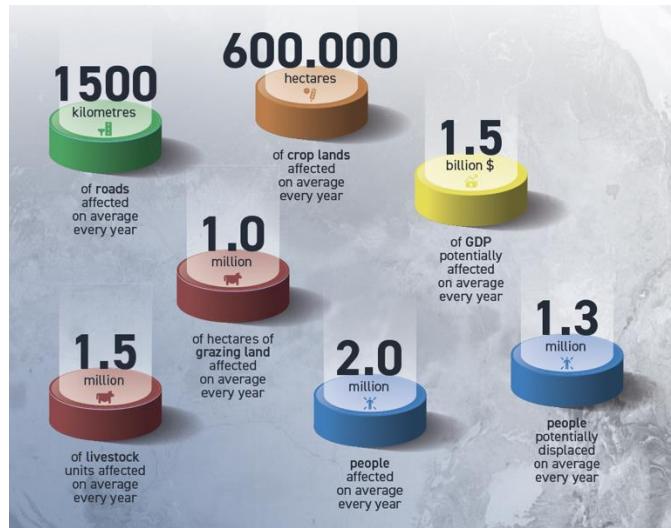
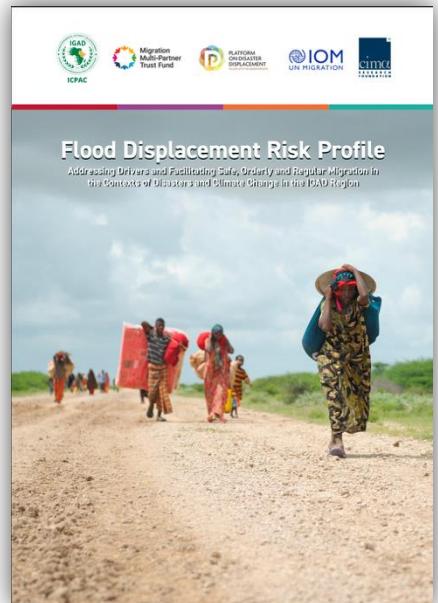


# OUTPUT 2

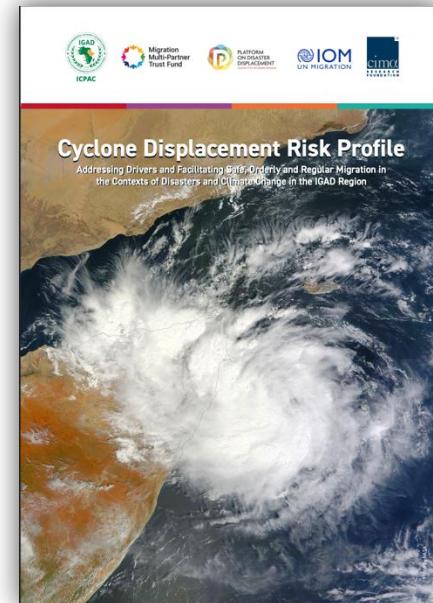
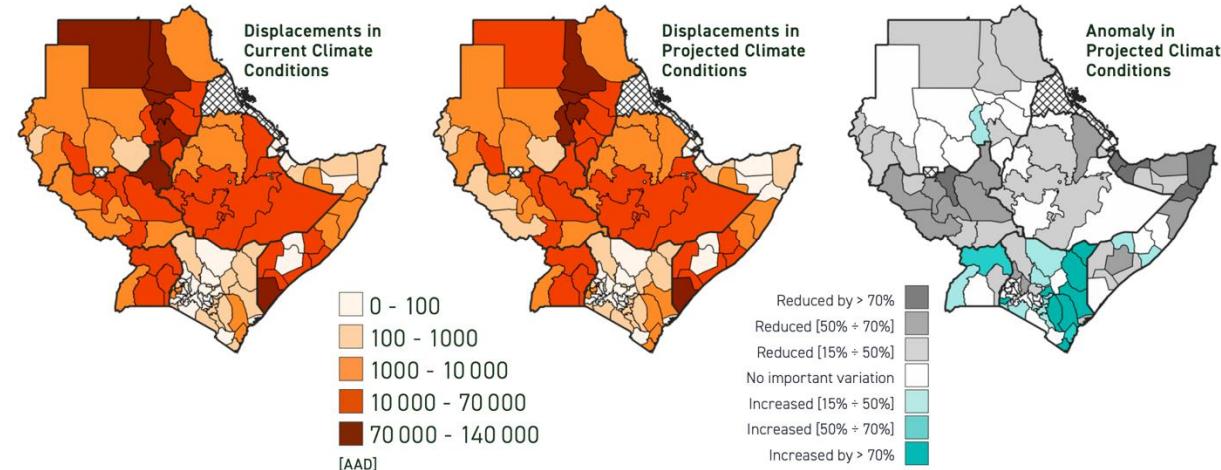
- **Early Warning Systems:** ICPAC develops displacement trend models (floods, droughts), integrates them into East African Hazards Watch, and improves displacement data governance.
- **Capacity & Communication:** Trainings for Member States and ICPAC, co-designed early warning bulletins, human mobility webpage, and engagement in regional/global knowledge platforms.
- **Preparedness:** Review and update preparedness plans at local level and design a regional model plan for cross-border climate mobility.



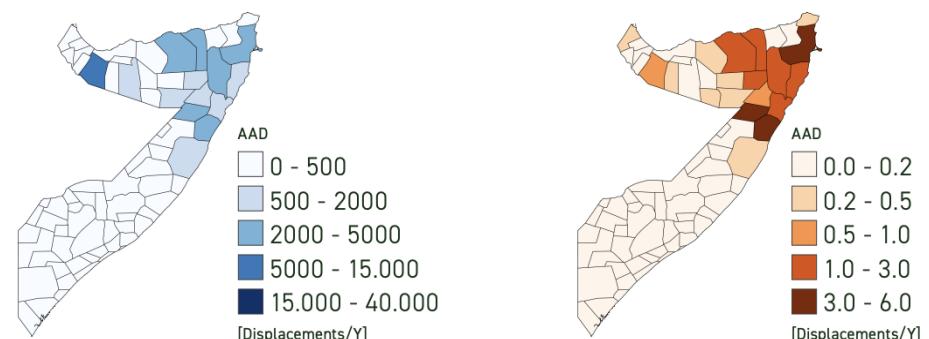
# TWO DISPLACEMENT RISK PROFILES DEVELOPED



ANNUAL AVERAGE DISPLACEMENT TRIGGERED BY FLOOD EVENTS AT PROVINCE OR COUNTY LEVEL



ANNUAL AVERAGE NUMBER OF PEOPLE POTENTIALLY DISPLACED BY CYCLONE EVENTS AT DISTRICT LEVEL (ON THE LEFT) AND THE CORRESPONDING PERCENTAGE IN RESPECT TO THE TOTAL POPULATION AT DISTRICT LEVEL (ON THE RIGHT)



The maps above show the annual average number of people potentially displaced due to cyclone events at district level: the total amount at country scale is around 30000, approximately 1/6 of the affected population. The northeastern part of the country is the one suffering the most.

# MODELLING APPROACHES

- Short-term visualization: historic mobility & climate data.
- Long-term predictive analytics:
  - Machine learning for seasonal mobility.
  - Agent-Based Modelling (ABM).
- Hybrid ABM-ML Model
 

Combines Agent-Based Modeling (ABM) behavioral mechanisms with Machine Learning predictive strength

  - Systems dynamics models.

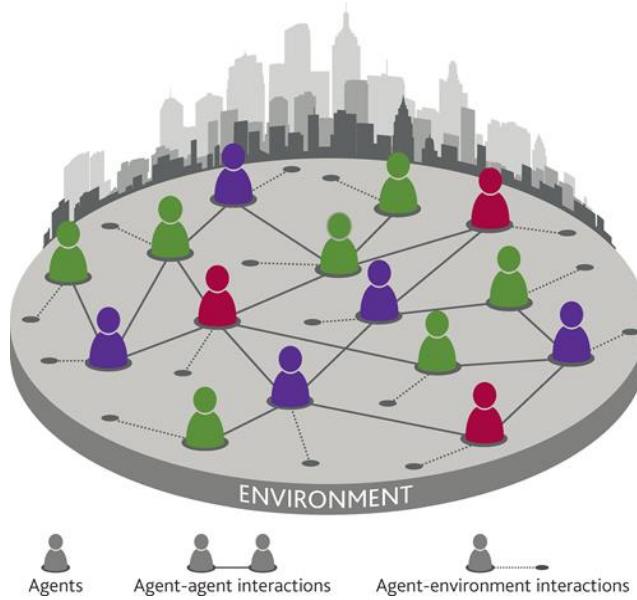
## Results of ABM modelling-AN EXAMPLE



# **FLOOD DISPLACEMENT MODELLING**

# DECISION-MAKING MODEL - AGENT-BASED MODEL

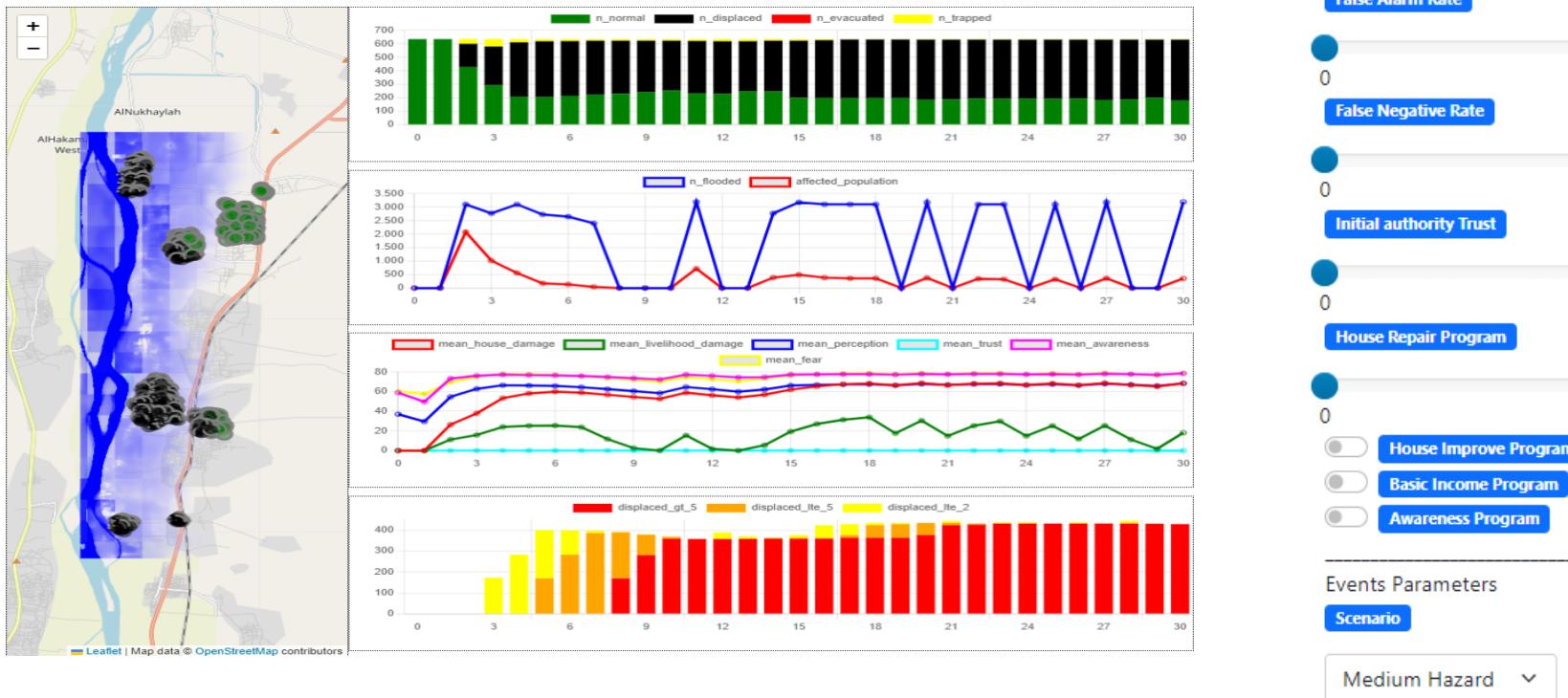
*An ABM is “a computational method for simulating the actions and interactions of autonomous decision-making entities in a network or system, with the aim of assessing their effects on the system as a whole.” (Dawson et al., 2011)*



## DECISION-MAKING MODEL - AGENT-BASED MODEL

- Model flood displacement using **systems perspective**
- Consider **dynamic** exposure and vulnerability
- Consider **local characteristics** and **cultural elements**
- Understand displacement **risk drivers**
- Provide displacement **policy analysis**
- **Support decision-making**

# SCENARIO SIMULATION



## MAIN INSIGHTS FOR THE MODEL

EWS + Basic Income Programme BI is the best investment if the combination of more policies is not possible.

House Repair Programme reduces displacement duration, but it increases the n°

of displaced HHs. HRP coupled with a Build Back Better Programme (BBB )to also reduce structural vulnerability.

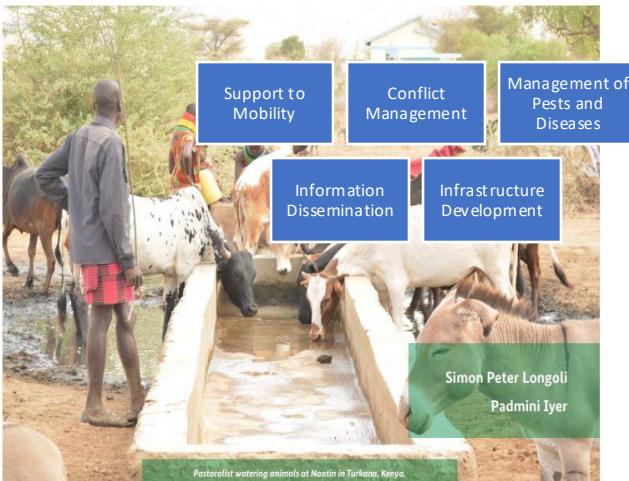
Prioritize scalable plans, according to governments' resources.

Communities' needs and characteristics at the core for crafting strategic interventions.

Long-term effects have to be considered.

# DROUGHT DISPLACEMENT MODELLING

# GIZ HMCCC



- Objectives:

- Determine the impacts of droughts on the livelihoods of pastoral and agro-pastoral communities;
- Analyzing migration decisions in response to the effects of droughts;
- Explore activities and policies that would increase the resilience to droughts.

- Findings:

- Climate change impacts livelihoods
- Declining animal health and livestock loss

There are norms that govern HMCCC

Social norms

Governmental regulations

HMCCC interacts with conflict dynamics

Insecurity exacerbates the impacts of climate change

Immobility

Likelihood of conflict is increased

Gender

Women participate in hazardous low return activity

Men assume primary responsibility for decisions around migration

# MODELLING APPROACH (1): MACHINE LEARNING AND REGRESSION ANALYSIS

- Utilise machine learning to explore the potential to develop a seasonal forecast of pastoralist displacement across international borders
- **Uses data detailing historic:**
  - Environmental and climate events / change
  - Records of historic mobility patterns
  - Other data around influences on drivers of mobility – for example, conflict, disease, market prices
  - (also risk profiling of vulnerable households)
- Aims to identify **relationships between climate, mobility and other influences**

## INDICATOR / PROJECTION:

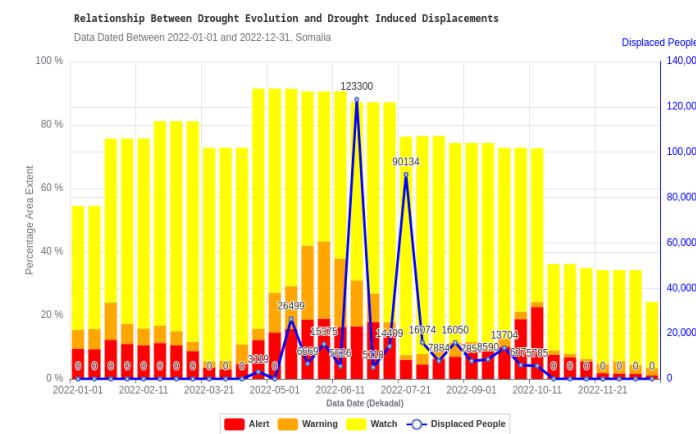
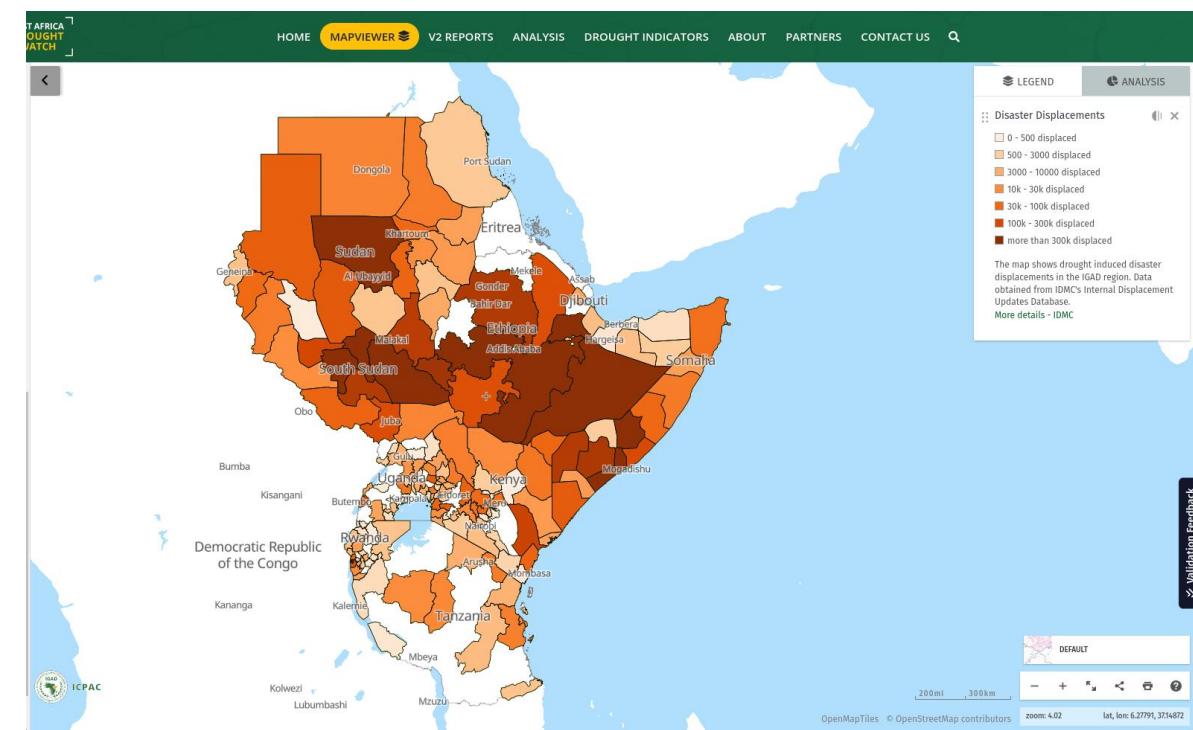
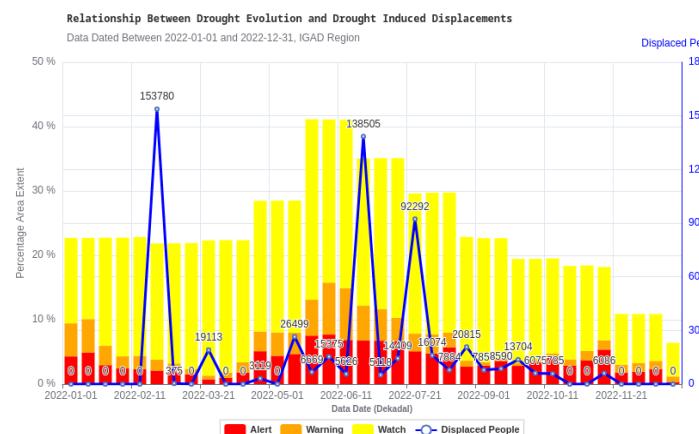
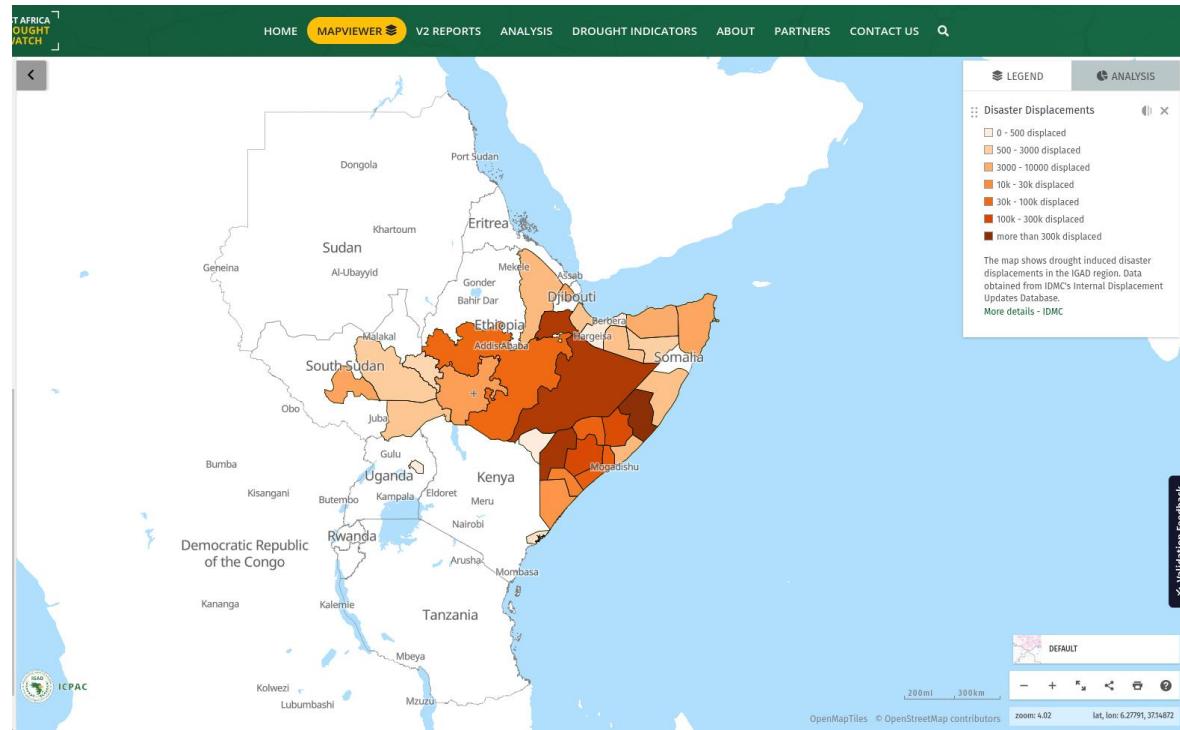
- Provide an indicator of the conditions under which we would expect to see waves of migration in the future

## MODELLING APPROACH (2): IDMC SYSTEMS DYNAMIC MODEL

- Utilise existing IDMC systems dynamic model to understand policy options to reduce displacement pressure
  - **Uses data detailing historic:**
    - Normalised Difference Vegetation Index (NDVI) data
    - Livestock population
    - Livestock prices
    - Displacement
    - Internally Displaced Population
  - Aims to identify **relationships between pasture, livestock population, livelihood loss, displacement**
  - **SCENARIOS – under what conditions...**
    - Provides an opportunity to look at relationships between conditions associated with livelihood loss and displacement, and the potential policy interventions that can be made

# **DATA COLLECTION AND SYSTEM INTEGRATION FOR OPERATIONAL USE**

# DISPLACEMENT DATA INTEGRATION WITH DROUGHT WATCH



# HUMAN MOBILITY WATCH(DEMO)

10/28/25, 8:24 AM

igad-dashboard

## System Architecture & Real-time Data Integration

### IDMC Global Database

**Purpose:** Internal displacement figures  
**API:** helix-tools-api.idmcdbs.org  
**Data:** IDP counts, events, analysis  
**Update:** Daily  
`GET /external-api/gidd/displacement-data?country=ETH&year=2024`

### IOM DTM

**Purpose:** Mobility & flow tracking  
**API:** dtm.iom.int/api/v1  
**Data:** Cross-border flows, sites  
**Update:** Weekly/Monthly  
`GET /api/v1/flow-monitoring?country=SM&type=cross_border`

### ICPAC Climate

**Purpose:** Climate triggers (CDI)  
**API:** droughtwatch.icpac.net  
**Data:** Drought indicators, forecasts  
**Update:** Dekadal (10-day)  
`GET /api/drought-indicators/cdi?region=eastern_africa&format=json`

### HDX Platform

**Purpose:** Humanitarian datasets  
**API:** data.humdata.org/api/3  
**Data:** Multi-source aggregation  
**Update:** Variable by source  
`GET /3/action/package_search?q=displacement+igad`

## System Capabilities & Impact

### Real-time Monitoring

- Live displacement tracking across IGAD
- Climate trigger early warnings
- Cross-border movement flows
- Humanitarian needs assessment
- Mobile alerts to at-risk communities

### Predictive Analytics

- 15-year displacement projections
- Climate hotspot identification
- Seasonal pattern recognition
- Early warning integration
- Risk scenario modeling

### Decision Support

- Policy recommendation engine
- Resource allocation optimization
- Multi-stakeholder dashboards
- Impact assessment tools
- Intervention effectiveness tracking

**80%**  
Displacement reduction potential

**4.3M**  
People protected by 2040

**8**  
IGAD countries covered

**24/7**  
Real-time monitoring

### IGAD Member States

Djibouti  
Ethiopia  
Somalia  
Sudan  
Eritrea  
Kenya  
South Sudan  
Uganda

### Data Partners

- Internal Displacement Monitoring Centre
- IOM Displacement Tracking Matrix
- ICPAC Climate Prediction Centre
- OCHA Humanitarian Data Exchange
- World Bank Groundswell Project

### Applications

- Humanitarian response planning
- Climate adaptation strategies
- Migration policy development
- Early warning systems
- Regional cooperation frameworks

### Technical Stack

- React.js + Tailwind CSS
- Recharts visualization library
- RESTful API Integration
- Real-time data processing
- Responsive design

IGAD Climate-Induced Human Mobility and Displacement Monitoring System | Last Updated: 10/28/2025 | Integrating IDMC, DTM, ICPAC & HDX APIs for comprehensive regional monitoring

localhost:3000

2/2

10/28/25, 8:25 AM

igad-dashboard

## East Africa Mobility Watch

Real-time monitoring of human mobility and displacement patterns in the context of climate change across the IGAD region

Country/Region

All IGAD Region

Time Period

2020-2024 (5 Years)

Displacement Type

All Causes

Export Report

Live Data Integration

Apply Filters

### Total Displaced

**2.9M**

Internal displacement  
 $\downarrow 11\%$  from 2023

### New Movements

**890K**

2024 new cases  
 $\uparrow 15\%$  from 2023

### Climate-Induced

**2.1M**

Drought & flood related  
75% of total

### Returns

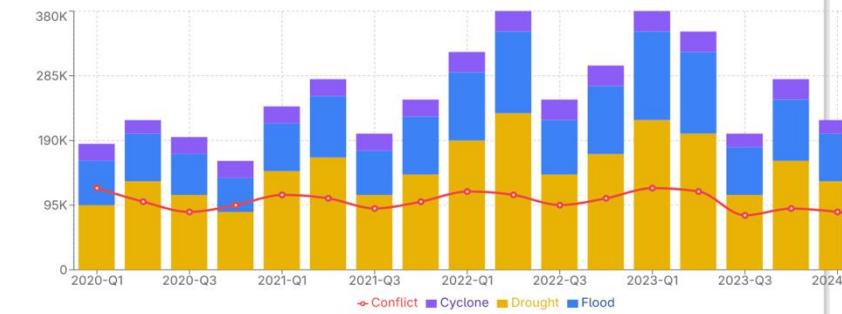
**450K**

People returned home  
 $\uparrow 25\%$  from 2023

Overview Trends Countries Mobility Projections

## Detailed Trend Analysis

### Displacement Causes by Quarter



### Dry Season (Dec-Mar)

### Peak Drought

Highest displacement from water scarcity

### Long Rains (Apr-Jun)

### Flood Risk

River flooding and displacement

### Short Rains (Oct-Nov)

### Recovery

Return movements increase

## System Architecture & Real-time Data Integration

### IDMC Global Database

**Purpose:** Internal displacement figures  
**API:** helix-tools-api.idmcdbs.org

### IOM DTM

**Purpose:** Mobility & flow tracking  
**API:** dtm.iom.int/api/v1

### HDX Platform

**Purpose:** Humanitarian datasets  
**API:** data.humdata.org/api/3

localhost:3000

1/2

## NEXT STEPS

- The need for early warning on risk of displacement due to drought, floods, landslides, conflict (later stage)...etc.
- Expand the evidence base on human mobility trends and risk of displacement in the IGAD region.
- Risk informed preparedness and contingency plans to better anticipate and manage mobility.
- Reaching last mile - (HUSIKA app) and strengthen local voices at global forums.
- Develop Human Mobility Watch as an interactive regional knowledge hub on mobility (with monitoring and forecasting capabilities) for early warning and decision-making support tool.



ICPAC

# Thank You!



MoDiAC



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